

Single Top Searches at the Tevatron

- ▶ Electroweak top quark production
- ▶ Signal and background modeling, b-tagging
- ▶ CDF analysis
 - ▶ s-channel, t-channel, combined results
- ▶ DØ analysis
 - ▶ Cut-based, Decision Trees, Neural Networks
- ▶ Cross section limits
- ▶ Outlook



Arán García-Bellido
on behalf of CDF and DØ

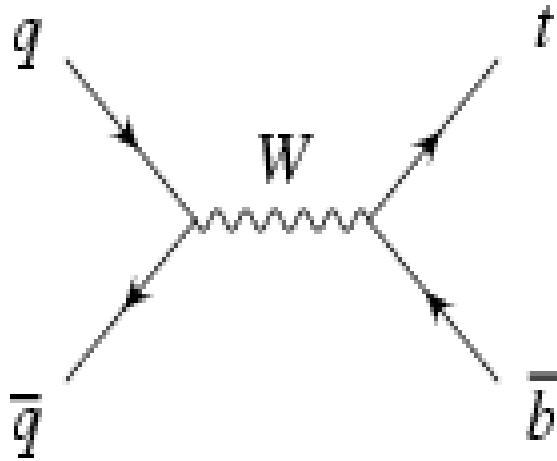


Top quark electroweak production

s-channel

TeV: 0.88 ± 0.11 pb

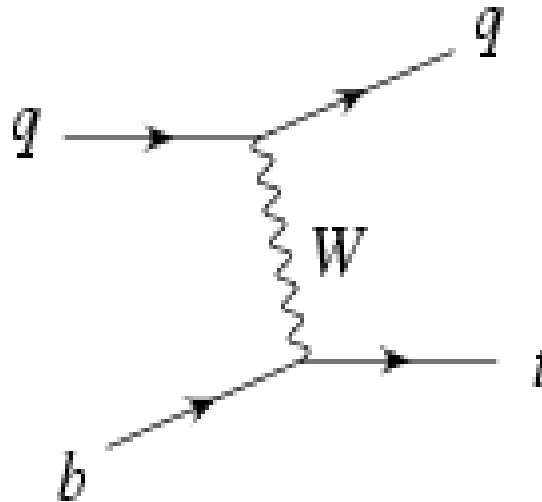
LHC: 10.6 ± 1.1 pb



t-channel

1.98 ± 0.25 pb

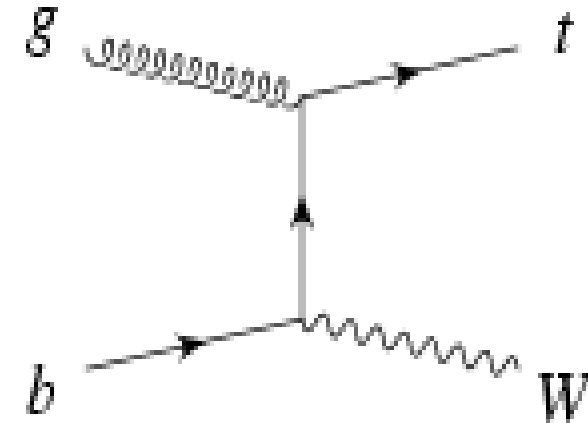
246.6 ± 0.25 pb



associated tW

< 0.1 pb

$62.0^{+16.6}_{-3.6}$ pb



Harris, Laenen, Phaf, Sullivan, Weinzierl: hep-ph/0207055
Cao and Yuan: hep-ph/0408180

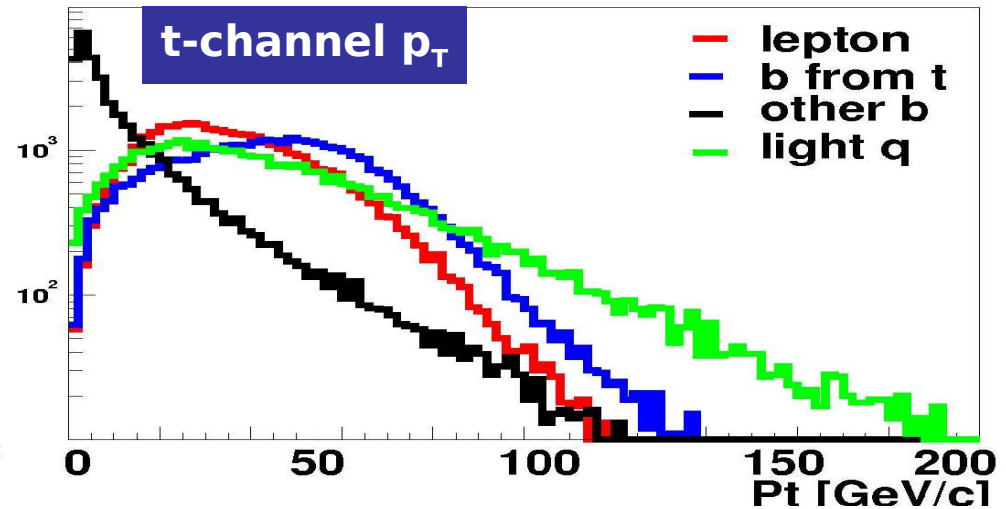
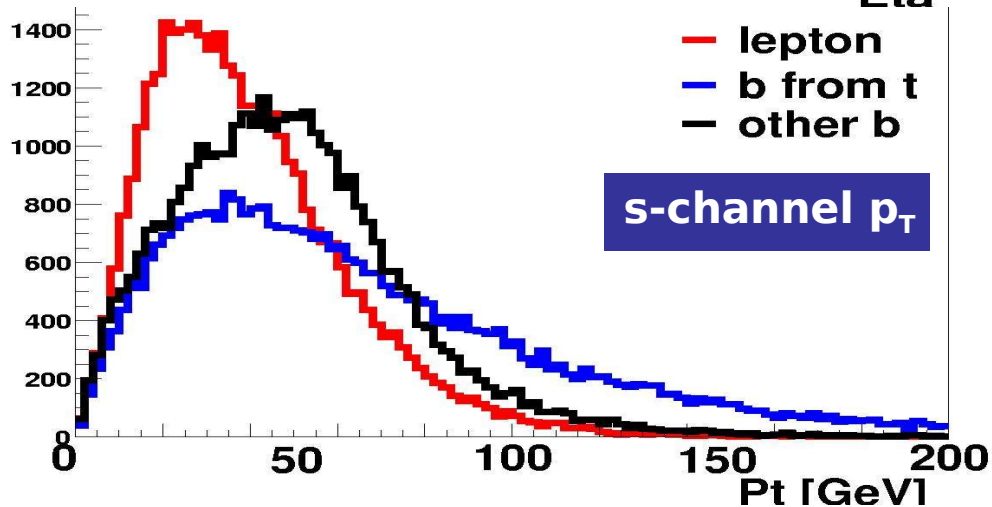
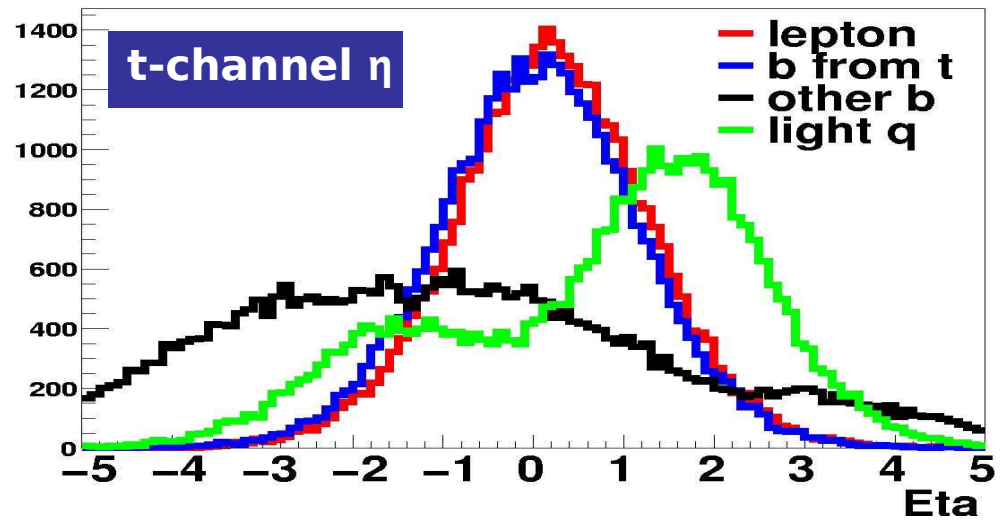
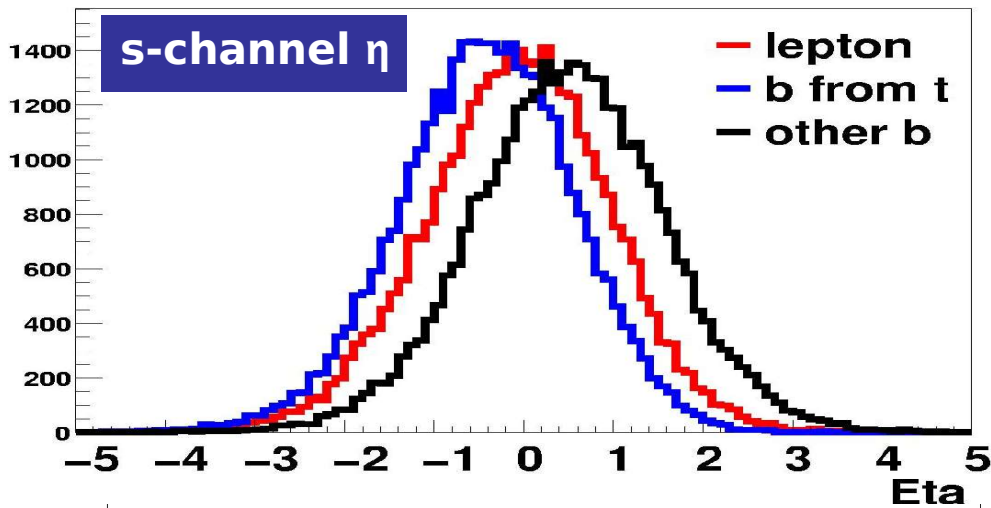
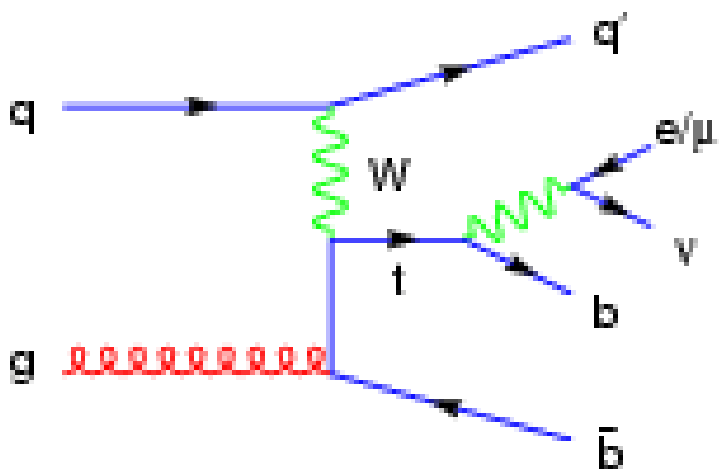
Tait: hep-ph/9909352
Belyaev, Boos: hep-ph/0003260

- ▶ Observe single top quark production in Run II ← **Final Goal!**
- ▶ Direct access to V_{tb} CKM matrix element
- ▶ Look for new physics: FCNC, 4th generation, W' , SUSY, ...
- ▶ Probe W+jets understanding and help SM higgs searches

Signal topology

We are looking for:

- ▶ One high p_T isolated lepton (from W)
- ▶ MET (ν from W)
- ▶ One b-quark jet (from top)
- ▶ A light flavor jet and/or another b-jet



Signal modeling

Have to get the t-channel right:

Avoid double counting when different diagrams produce same final states in different kinematic regions

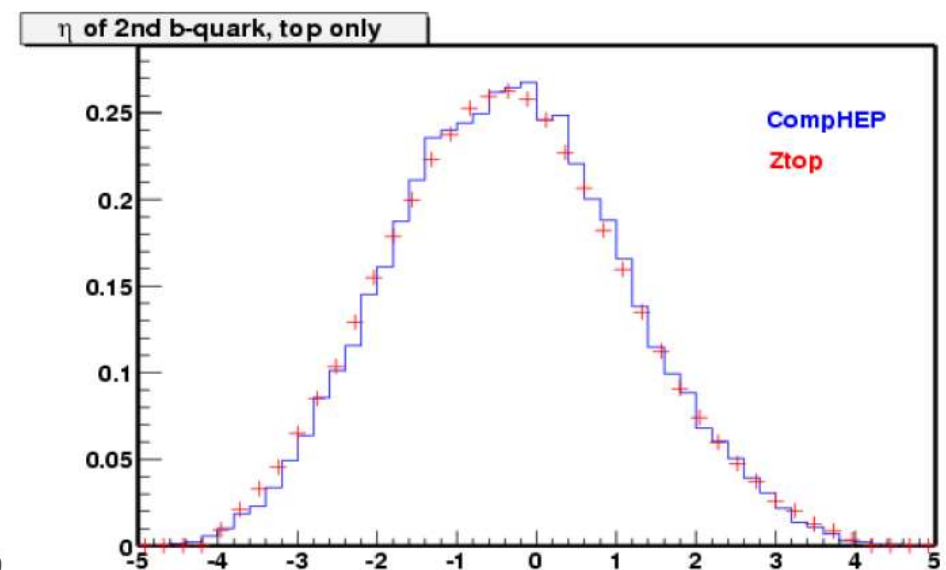
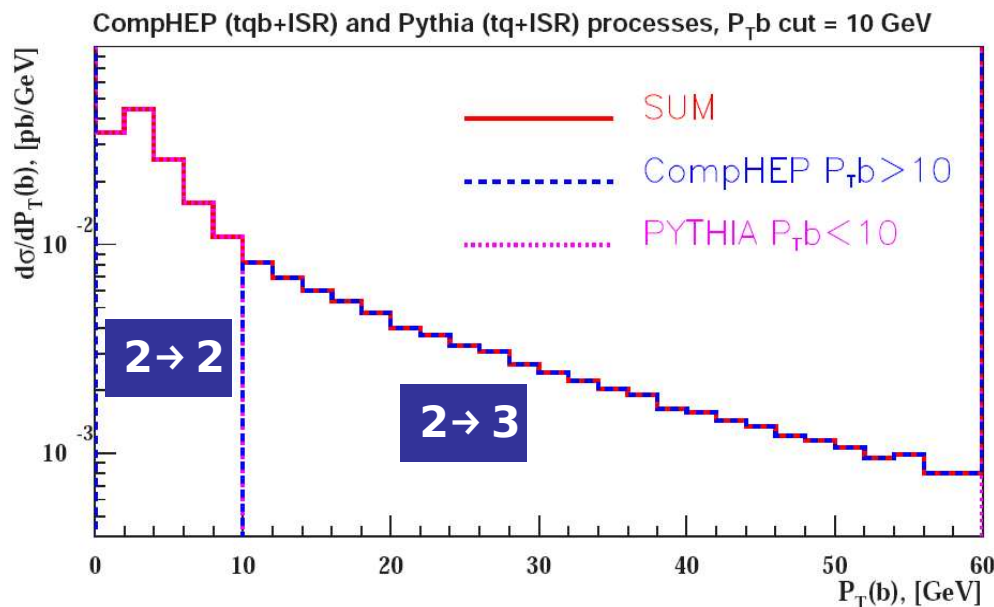
Use ZTOP as NLO benchmark <http://home.fnal.gov/~zack/ZTOP>

► CDF: Reweight MADEVENT to fit ZTOP

Generate: $b+q \rightarrow t+q'$ and $g+q \rightarrow t+b+q'$ separately and merge them to reproduce the b p_T spectrum from NLO

► DØ: “Effective” NLO CompHEP (also used in CMS)

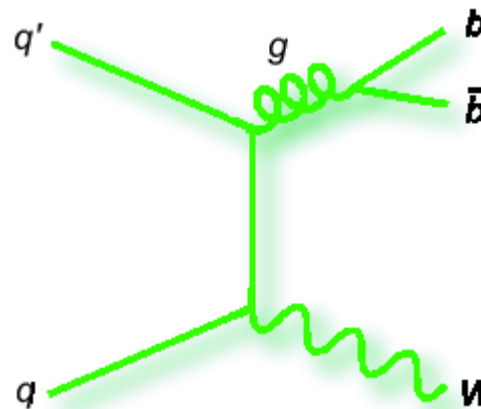
Match $2 \rightarrow 2$ and $2 \rightarrow 3$ processes using b p_T for cross over, normalize to NLO σ
Resulting distributions reproduce ZTOP & MCFM



Background modeling

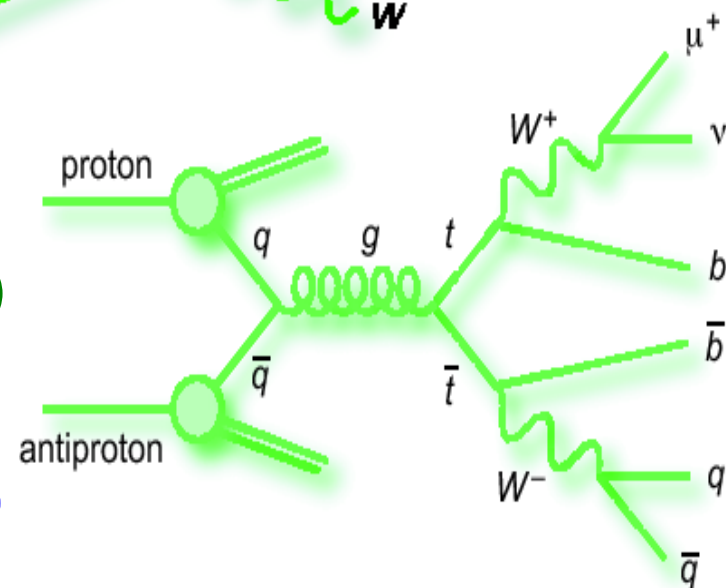
► W+jets

- Distributions from Alpgen
- Normalization from data
- Flavor fractions from Alpgen



► Top pairs

- Use Pythia (CDF) and Alpgen (DØ)



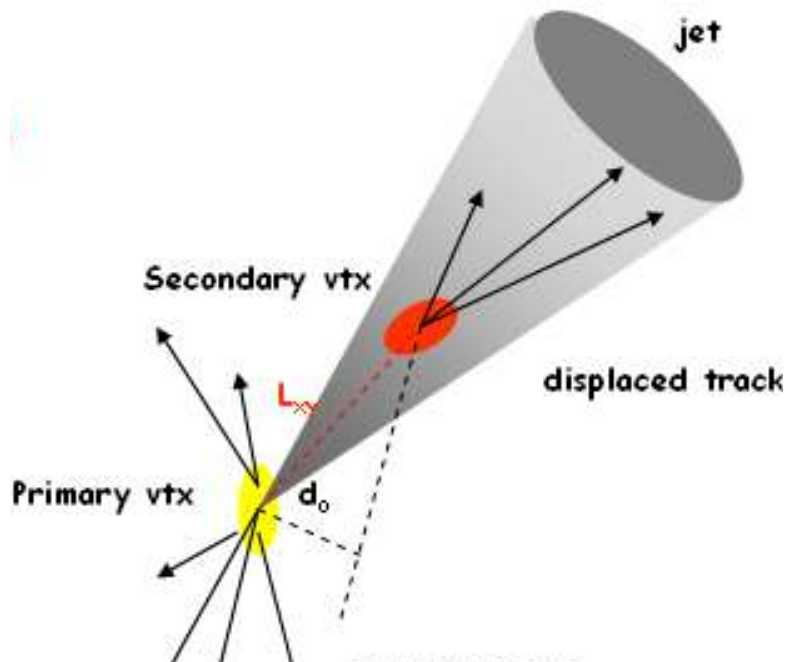
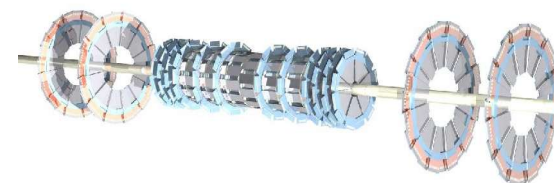
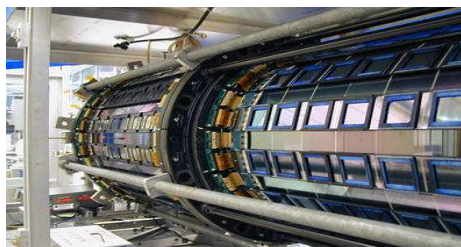
► Multijet events (misidentified lepton)

- From data

► Diboson (WZ, WW)

- Estimated from Pythia (CDF) and Alpgen (DØ)

Tagging b-jets



Both CDF and DØ use a secondary vertex reconstruction algorithm

~40% b-jet efficiency

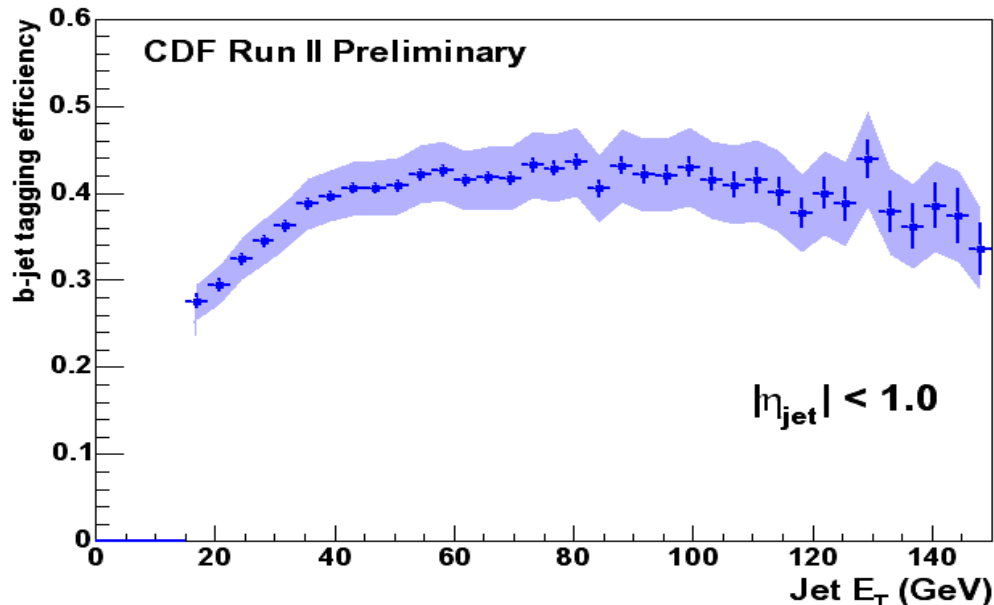
~ 1% mistag efficiency (light jets)

Charm tagging rate ~30% of b

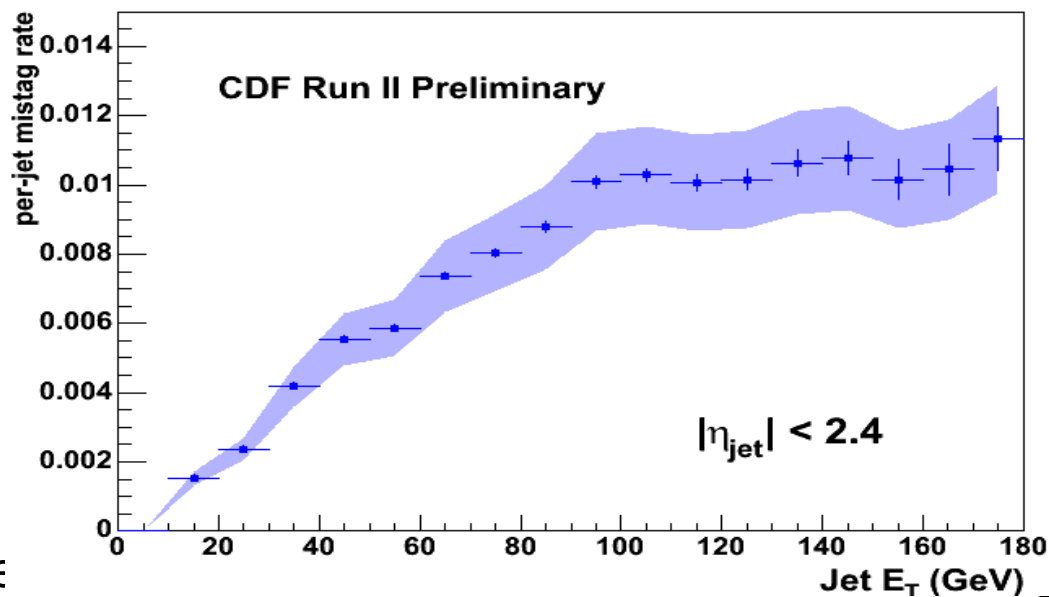
► DØ performance is very similar

► CDF has recently released an improved version (15% better)

SecVtx B-tagging Efficiency ($t\bar{t}$ Events)



SecVtx Mistag Rate



CDF search strategy

Phys.Rev.D71, 012005 (2005)

Lepton(e, μ): $p_T > 20$ GeV, $|\eta| < 1.0$

Jets: $N_{\text{jets}} = 2$, $E_T > 15$ GeV, $|\eta| < 2.8$

MET: $\text{MET} > 20$ GeV

Other clean-up cuts

b-tagging: At least one SVT jet

m_t window: $140 < M(\ell\nu b) < 210$ GeV

s, t and combined searches

Joint likelihood for separate searches

Likelihood with H_T distribution for s+t

s-channel: use two tagged events

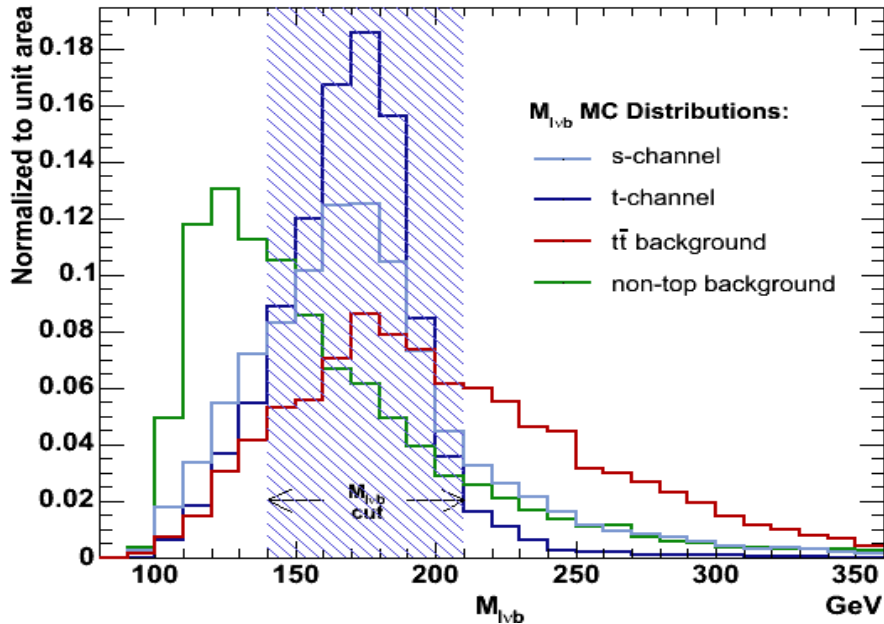
t-channel: use 1 tag $Q \times \eta$ distribution

Combined (s+t): use H_T distribution



CDF analysis after selection

CDF Run II Preliminary



Process	Combined	1-tag	2-tag
t -channel	2.8 ± 0.5	2.7 ± 0.4	0.02 ± 0.01
s -channel	1.5 ± 0.2	1.1 ± 0.2	0.32 ± 0.05
$t\bar{t}$	3.8 ± 0.9	3.2 ± 0.7	0.60 ± 0.14
Nontop	30.0 ± 5.8	23.3 ± 4.6	2.59 ± 0.71
Total background	33.8 ± 5.9	26.5 ± 4.7	3.19 ± 0.72
Total expected	38.1 ± 5.9	30.3 ± 4.7	3.53 ± 0.72
Observed	42	33	6

Systematics

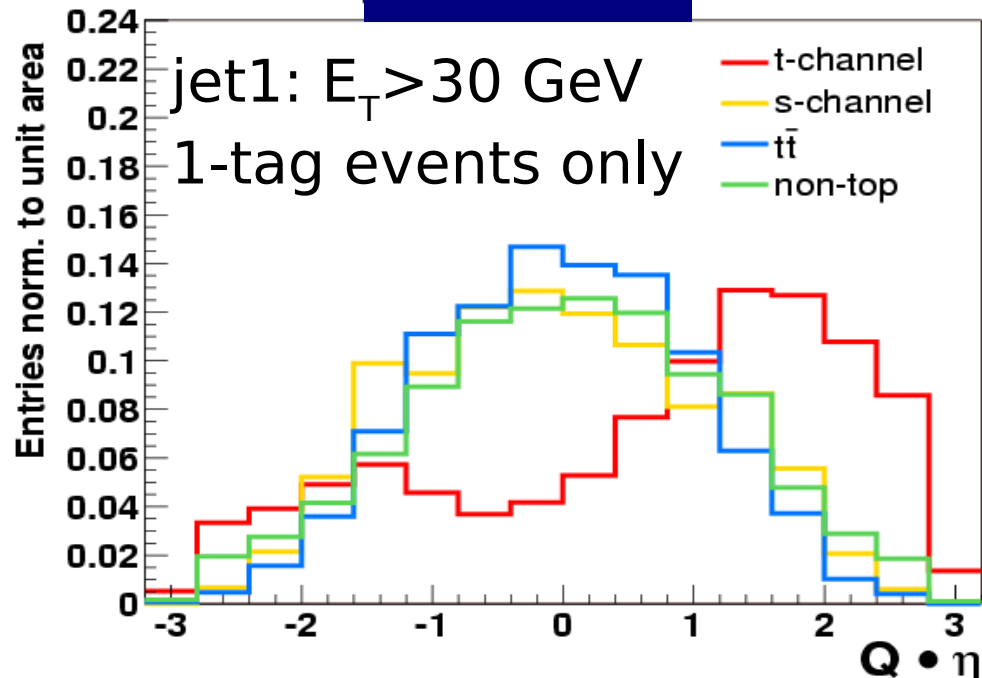
		t -channel	s -channel	Combined
1	JES	+2.4 -6.7	+0.4 -3.1	+0.1 -4.3
2	ISR	± 1.0	± 0.6	± 1.0
3	FSR	± 2.2	± 5.3	± 2.6
4	PDF	± 4.4	± 2.5	± 3.8
5	Generator	± 5	± 2	± 3
6	Top quark mass	+0.7 -6.9	-2.3	-4.4
7	$\epsilon_{\text{tag}}, \epsilon_{\text{ID}}, \text{luminosity}$	± 9.8	± 9.8	± 9.8

- Acceptance:
 $1.06 \pm 0.08\%$ (s -channel)
 $0.89 \pm 0.07\%$ (t -channel)

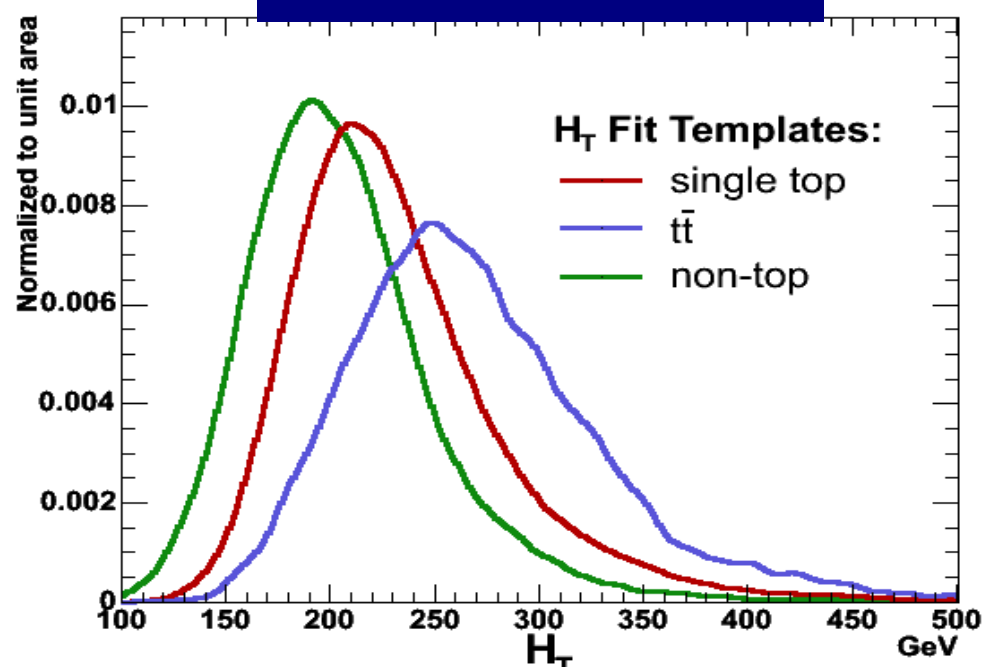
- Main systematics:
 B-ID 7%
 Luminosity 6%
 m_t 4%
 JES 4%

CDF distributions

CDF Run II **t-channel**

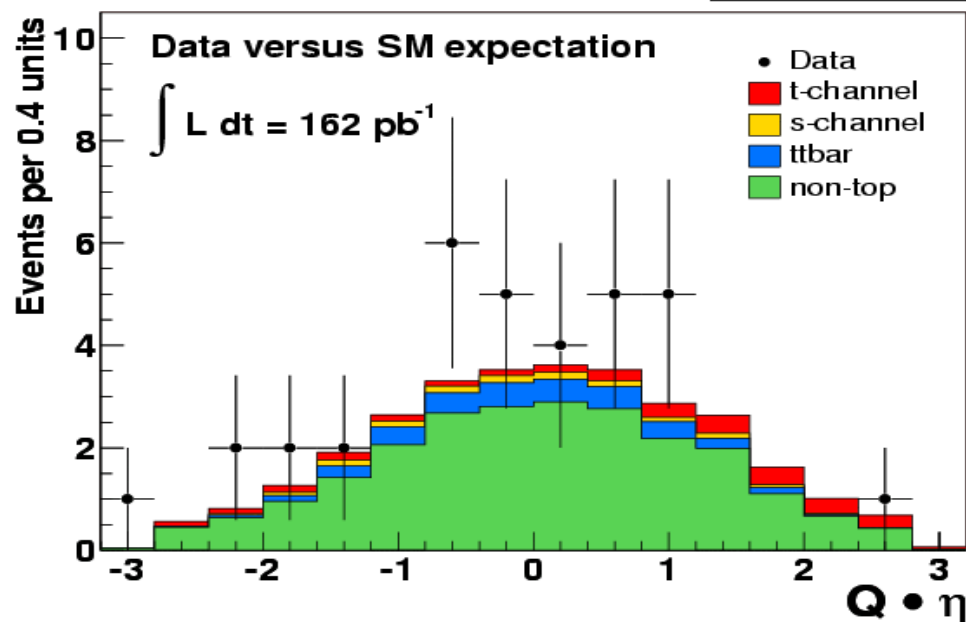


Combined search



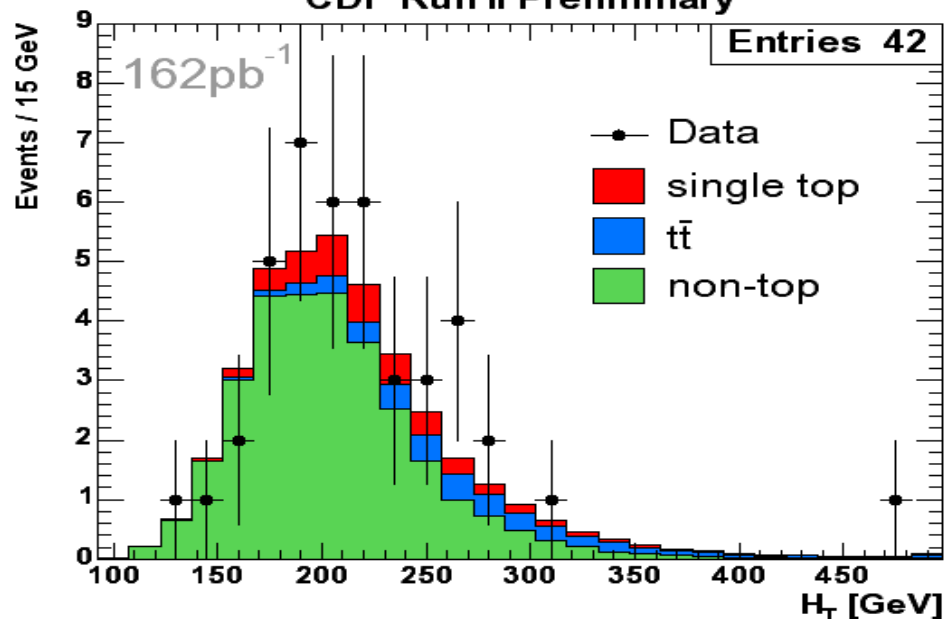
CDF Run II Preliminary

Entries 33



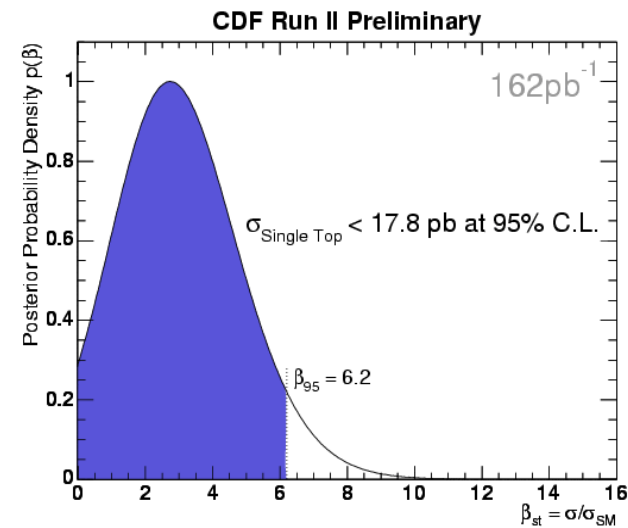
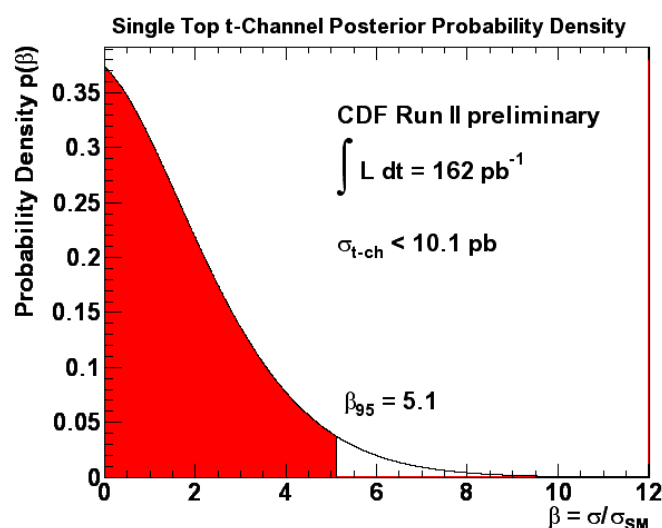
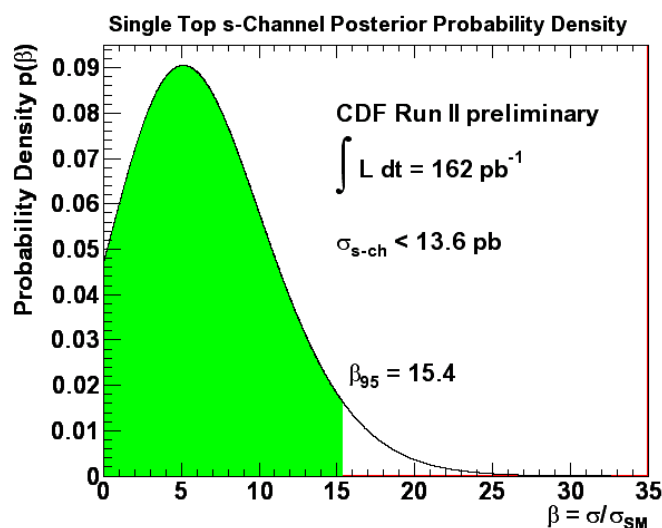
CDF Run II Preliminary

Entries 42



CDF limits

- ▶ Build a likelihood as a function of $\beta_i = \sigma_i / \sigma_i^{\text{SM}}$
- ▶ Background is allowed to float but is constrained to expectation
- ▶ Find maximum of the likelihood:
 - $L(\text{separate}) = L(2\text{tag events}) \otimes L(1\text{tag } Q \times \eta \text{ distribution})$
 - $L(\text{comb}) = L(H_T \text{ distribution})$
- ▶ Shape of systematic uncertainties included in the likelihood



Cross section upper limits (pb) at 95% CL: $\mathcal{L} = 162 \text{ pb}^{-1}$

s-channel

t-channel

Combined

Observed
13.6

Expected
12.1

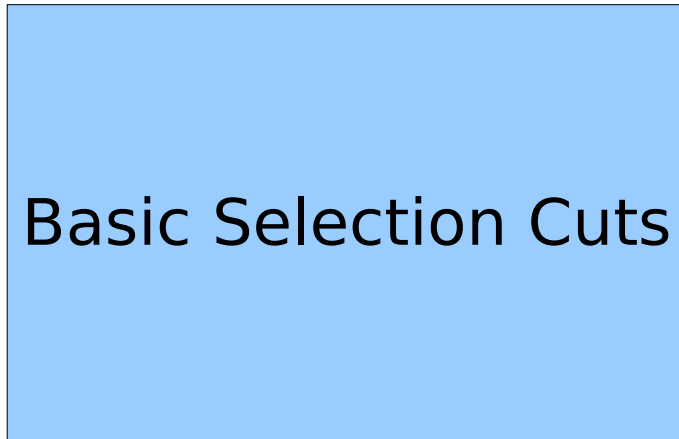
Observed
10.1

Expected
11.2

Observed
17.8

Expected
13.6

DØ search strategy



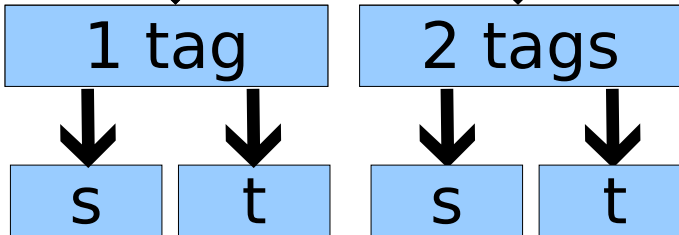
Lepton(e,μ): $p_T > 15$ GeV, $|\eta_{e(\mu)}| < 1.1$ (2.0)

Jets: $2 \leq N_{\text{jets}} \leq 4$, $E_T > 15$ GeV, $|\eta| < 3.4$

Jet1: $E_T > 25$ GeV

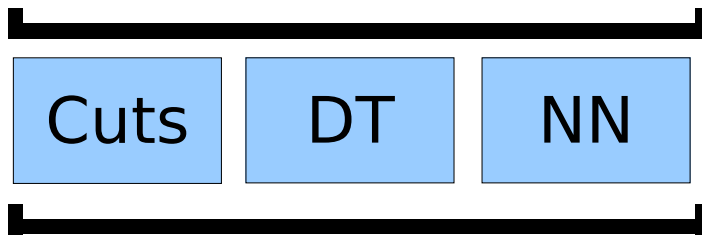
MET: MET > 15 GeV

Other clean-up cuts

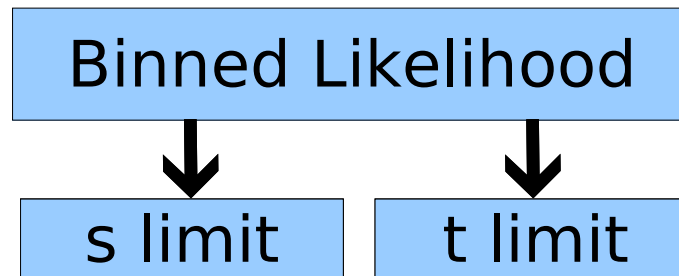


Require =1 and ≥ 2 SVT tags

t-channel: at least one non-b-tagged jet



Multivariate analysis: Simple cuts, Decision Trees, Neural Networks



Cut analysis: count events

DT, NN: use 2D output in a likelihood

DØ analysis after selection

- ▶ DØ optimizes selection to maximize acceptance
- ▶ Allows a lot of background at this stage!
- ▶ Then use multiple distributions to separate signal-background

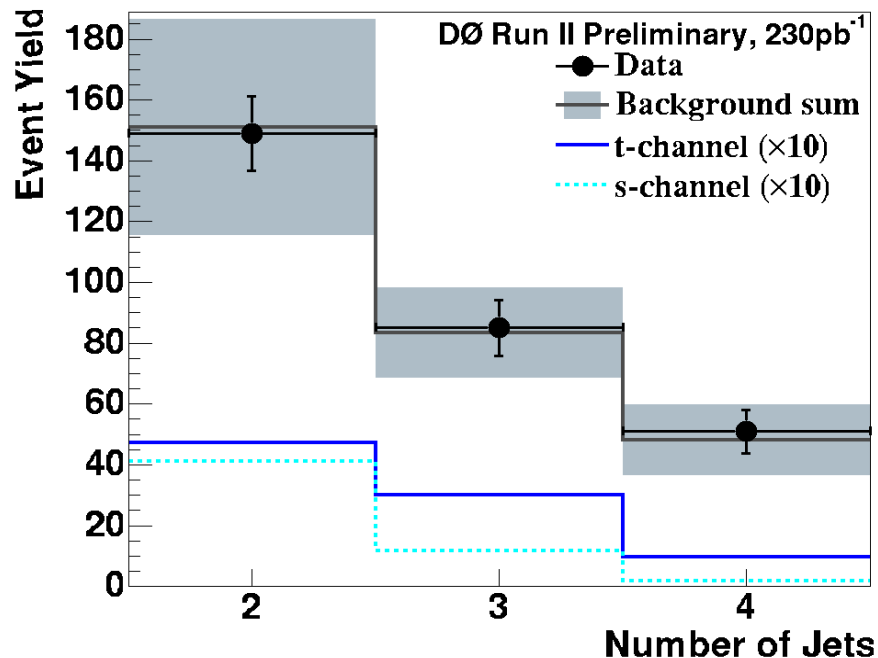
Source	<i>s</i> -channel search	<i>t</i> -channel search
<i>tb</i>	5.5 ± 1.3	4.7 ± 1.0
<i>tqb</i>	8.6 ± 1.9	8.5 ± 1.9
<i>t\bar{t}</i>	78.3 ± 18.3	75.9 ± 17.6
<i>W</i> +jets	169.1 ± 20.1	163.9 ± 18.7
Multijet	31.4 ± 3.3	31.3 ± 3.2
Total background	287.4 ± 43.6	275.8 ± 40.6
Observed events	283	271

- ▶ Acceptance: $2.7 \pm 0.2\%$ (*s*-channel) and $1.9 \pm 0.2\%$ (*t*-channel)
CDF has a more restrictive selection: optimize S/B to maximize separation and use one distribution

Systematic Uncertainties

Monte Carlo Systematic Uncertainties

Theory cross sections	15%
SVT modeling, single (double) tag	10%(20%)
Jet Energy Scale	10%
Trigger Modeling	6%
Jet Fragmentation	6%
Jet ID	5%
ℓ ID	5%



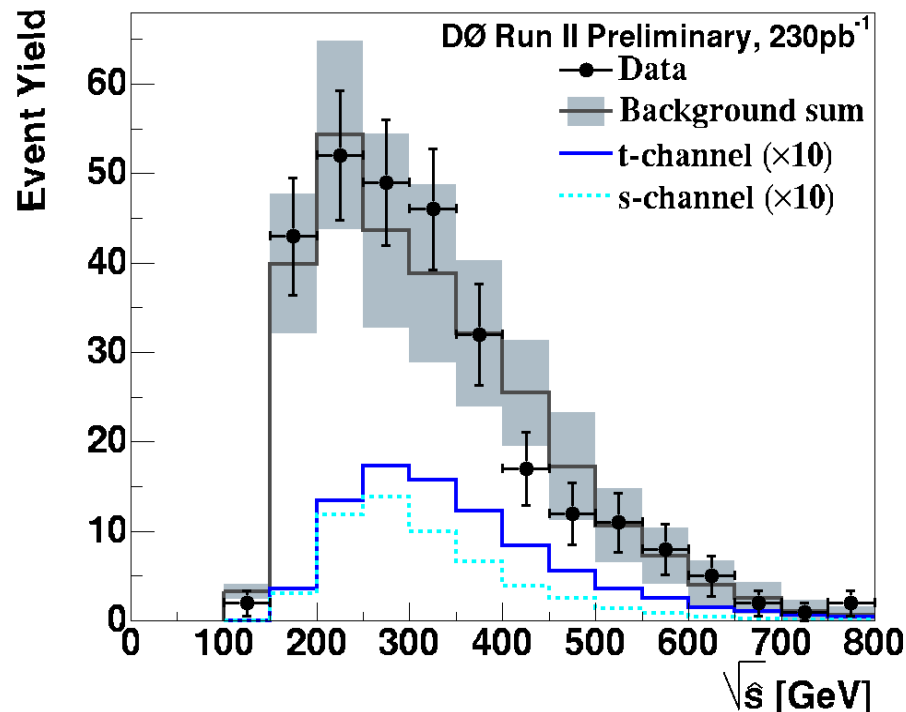
► Some systematic uncertainties also affect shape:

JES, b-tag and trigger modeling

► Total uncertainty:

	1 tag	2 tags
Signal acceptance	15%	25%
Background sum	10%	26%

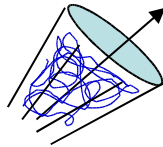
► Result is statistics limited



Discriminating variables

Individual object kinematics

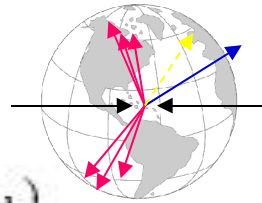
- $p_T(\text{jet1}_{\text{tagged}})$
- $p_T(\text{jet1}_{\text{untagged}})$
- $p_T(\text{jet2}_{\text{untagged}})$
- $p_T(\text{jet1}_{\text{nonbest}})$
- $p_T(\text{jet2}_{\text{nonbest}})$



This is where our phenomenology friends come so handy!

Global event kinematics

- $M_T(\text{jet1}, \text{jet2})$
- $p_T(\text{jet1}, \text{jet2})$
- $M(\text{alljets})$
- $H_T(\text{alljets})$
- $M(\text{alljets} - \text{jet1}_{\text{tagged}})$
- $H(\text{alljets} - \text{jet1}_{\text{tagged}})$
- $H_T(\text{alljets} - \text{jet1}_{\text{tagged}})$
- $p_T(\text{alljets} - \text{jet1}_{\text{tagged}})$
- $M(\text{alljets} - \text{jet}_{\text{best}})$
- $H(\text{alljets} - \text{jet}_{\text{best}})$
- $H_T(\text{alljets} - \text{jet}_{\text{best}})$
- $M(\text{top}_{\text{tagged}}) = M(W, \text{jet1}_{\text{tagged}})$
- $M(\text{top}_{\text{best}}) = M(W, \text{jet}_{\text{best}})$
- $\sqrt{\hat{s}}$



Three broad categories:

- ▶ Object kinematics
- ▶ Global event kinematics
- ▶ Angular correlations

Reconstruct W : from ℓ and ν

To reconstruct the top quark:

- ▶ s-channel: “best” jet algorithm

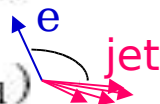
Chose the jet that gives m_t closest to 175GeV

- ▶ t-channel: lead b-tagged jet + W

Reconstruct q' : lead untagged jet

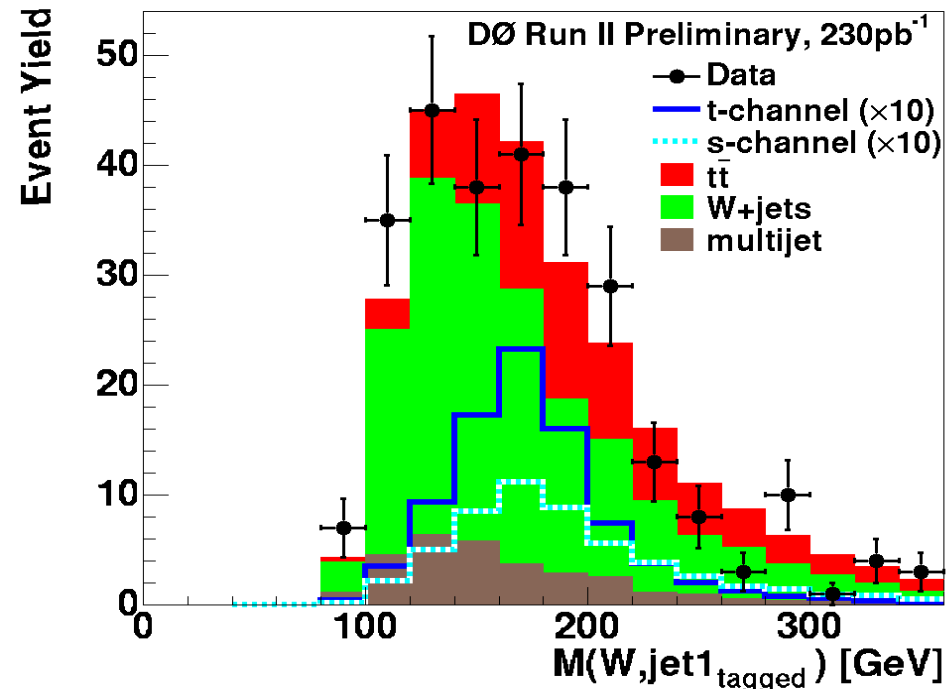
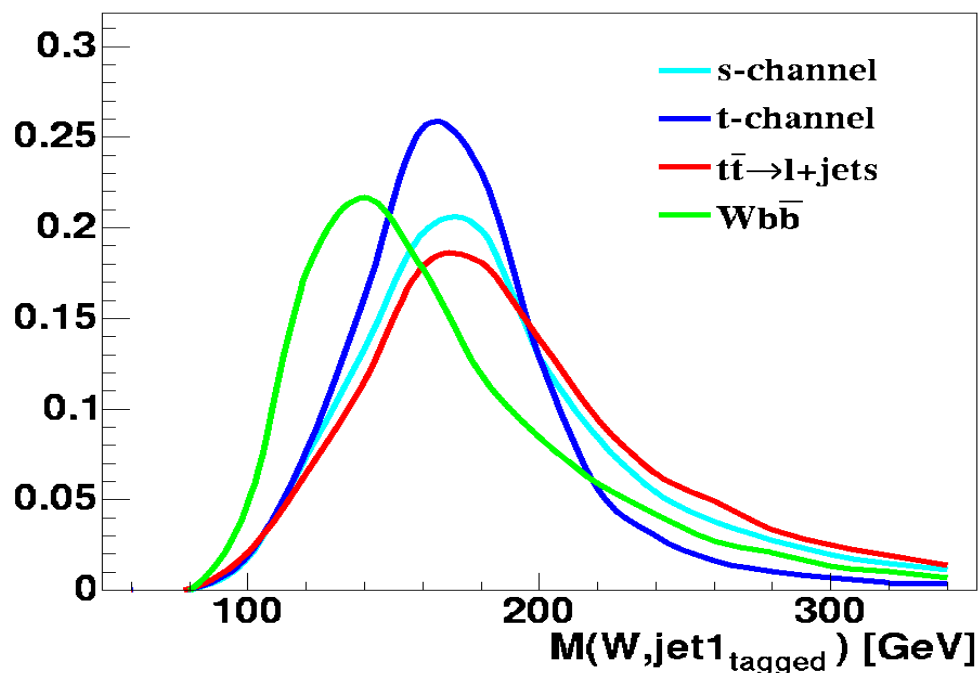
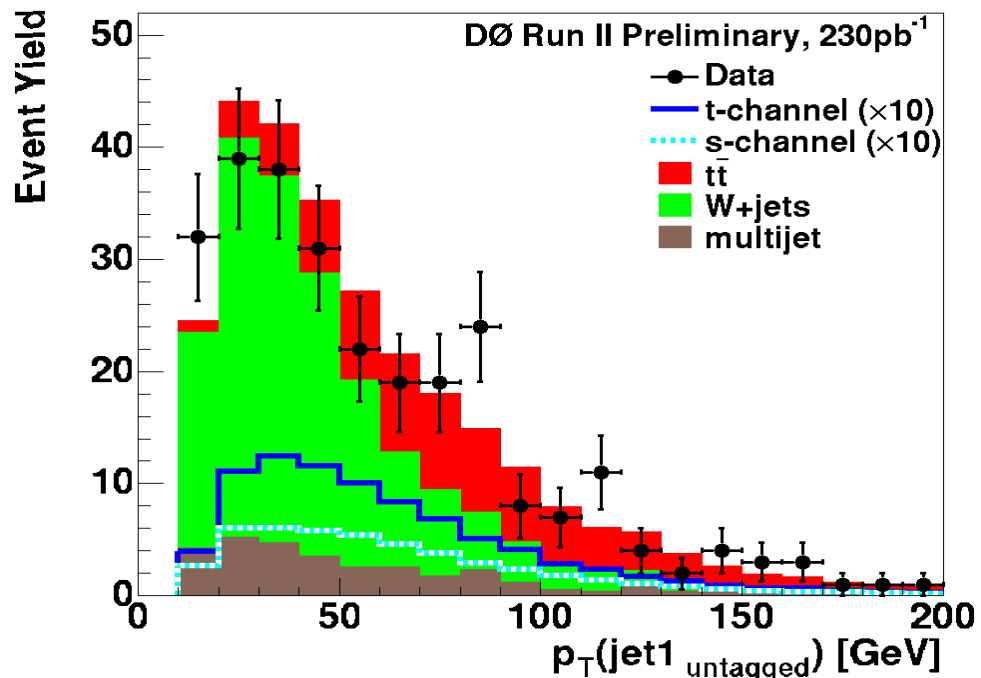
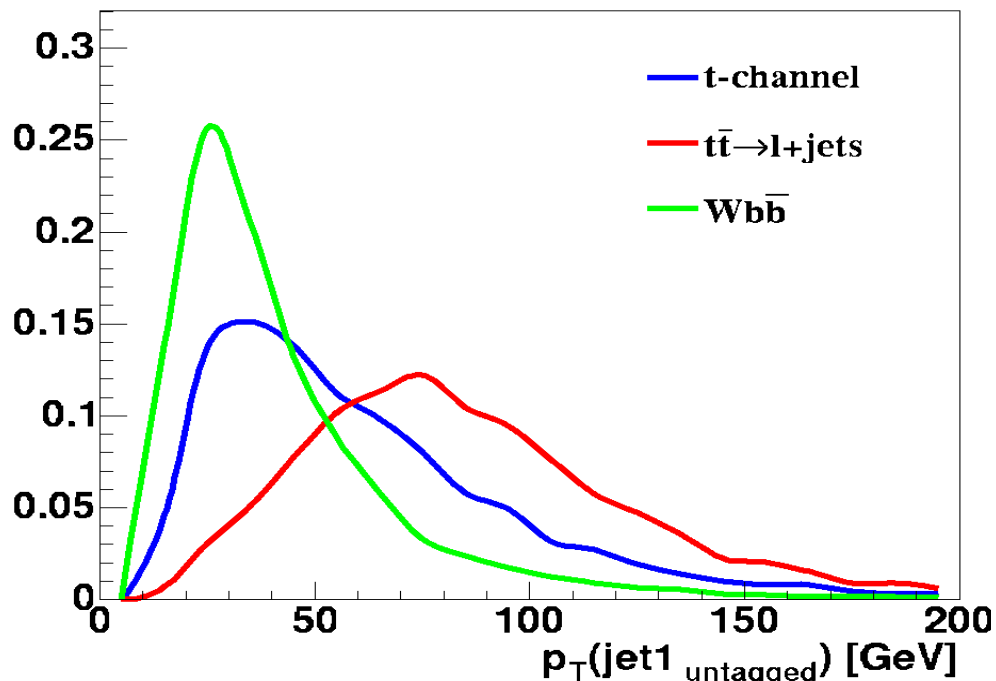
Angular variables

- $\Delta R(\text{jet1}, \text{jet2})$
- $Q(\text{lepton}) \times \eta(\text{jet1}_{\text{untagged}})$
- $\cos(\text{lepton}, Q(\text{lepton}) \times z)_{\text{top}_{\text{best}}}$
- $\cos(\text{lepton}, \text{jet1}_{\text{untagged}})_{\text{top}_{\text{tagged}}}$
- $\cos(\text{alljets}, \text{jet1}_{\text{tagged}})_{\text{alljets}}$
- $\cos(\text{alljets}, \text{jet}_{\text{nonbest}})_{\text{alljets}}$

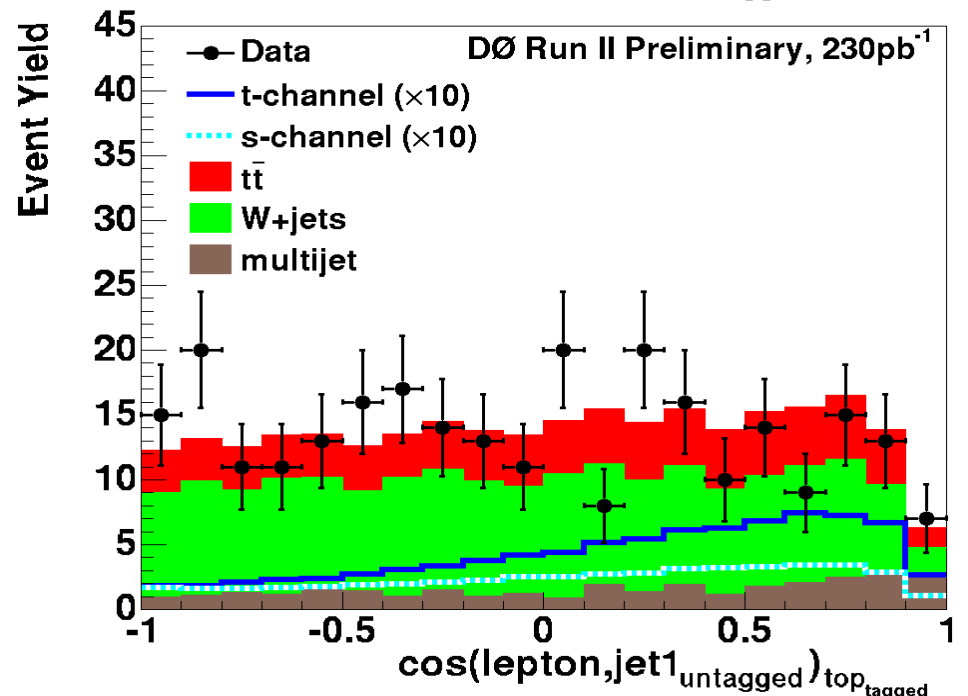
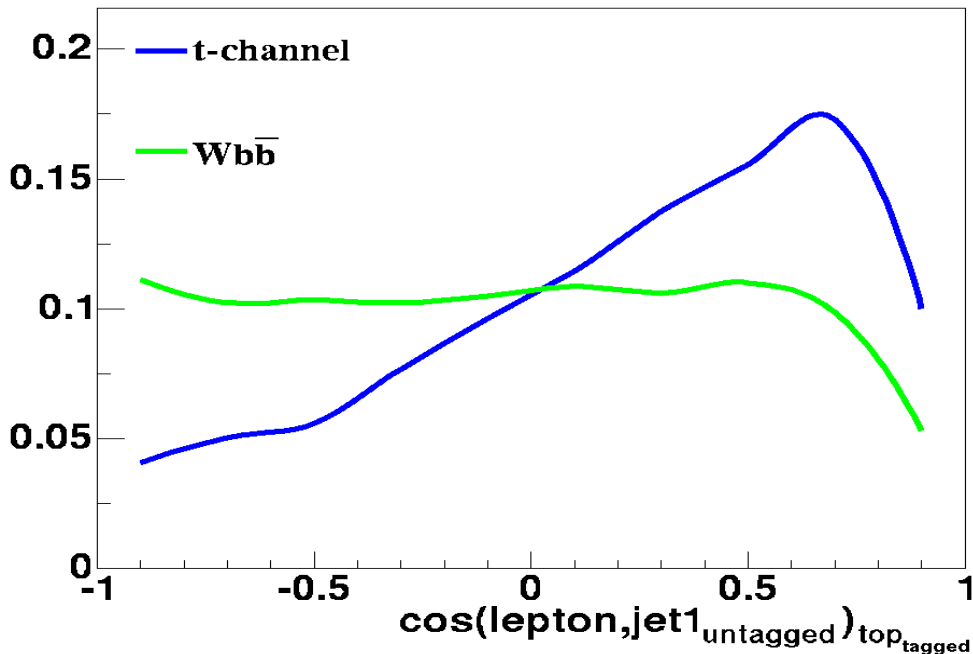
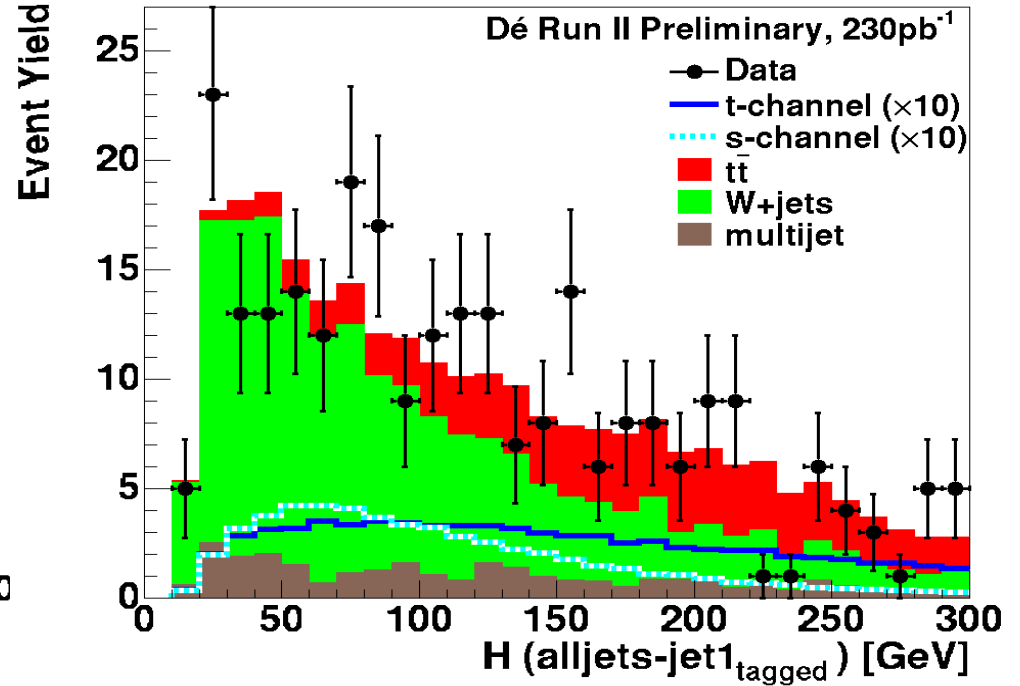
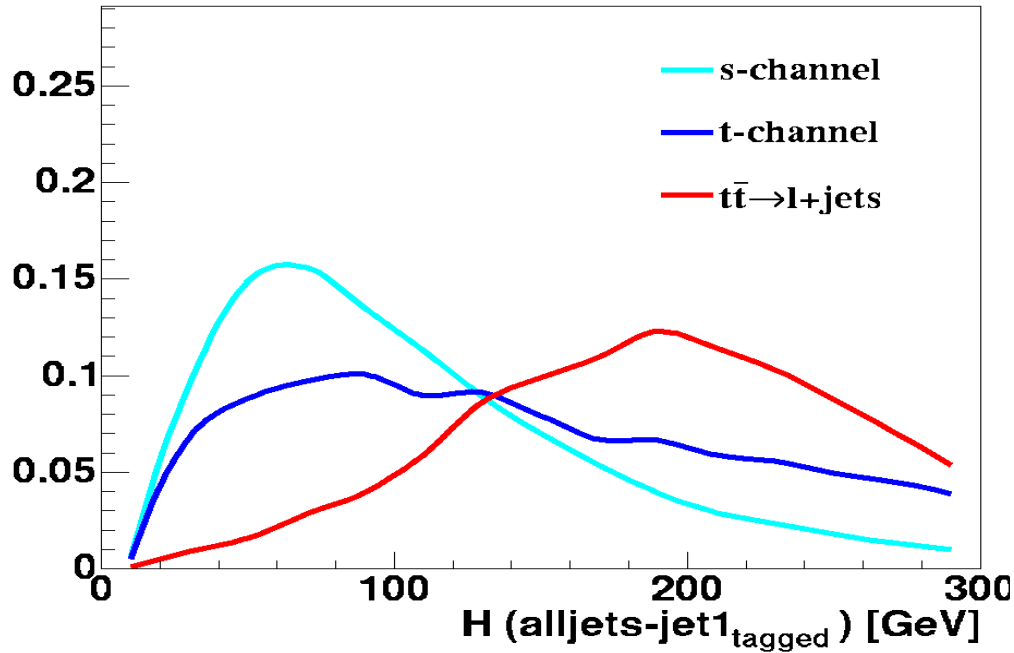


- s-channel search only
- t-channel search only
- used in both

Object kinematics



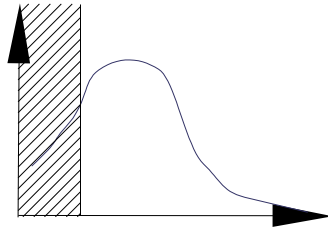
Event kinematics & angular variable



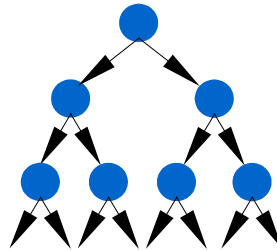
Separating signal from backgrounds

DØ has implemented three analysis methods:

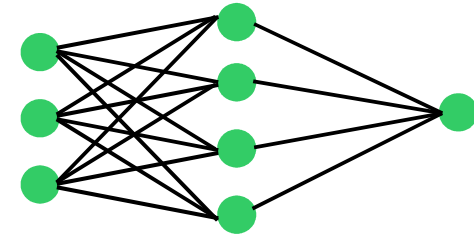
Cut-based



Decision Trees



Neural Networks



► Use same pool of discriminating variables for all 3 analyses

► Optimize separately for s-channel and t-channel

Input variables are independent of ℓ flavor, but optimize separately for e and μ datasets, given the different coverage and resolutions

► Focus on two dominant backgrounds: Wbb and $tt \rightarrow \ell + \text{jets}$

► A total of 8 sets of cuts/trees/networks:

$tb-Wbb$, $tb-tt \rightarrow \ell + \text{jets}$, $tqb-Wbb$ & $tqb-tt \rightarrow \ell + \text{jets}$ (for e and μ)

Cut-based analysis

- 1) Rate each discriminant variable according to its optimal cut
Optimal cut values are obtained minimizing the expected limit with respect to cut values derived from the signal sample
- 2) Try several sets of ANDed variables and re-optimize
- 3) Select the set that yields the lowest expected limit

	<i>s</i> -channel	<i>t</i> -channel
<i>tb</i>	4.5 ± 1.0	3.2 ± 0.8
<i>tqb</i>	5.5 ± 1.2	7.0 ± 1.6
<i>t\bar{t}</i>	27.6 ± 7.6	55.9 ± 12.3
<i>W</i> +jets	102.9 ± 13.7	72.6 ± 9.7
Mis-ID'd lepton	17.2 ± 2.0	17.0 ± 2.0
Background sum	153.1 ± 24.5	148.7 ± 24.8
Observed events	152	148

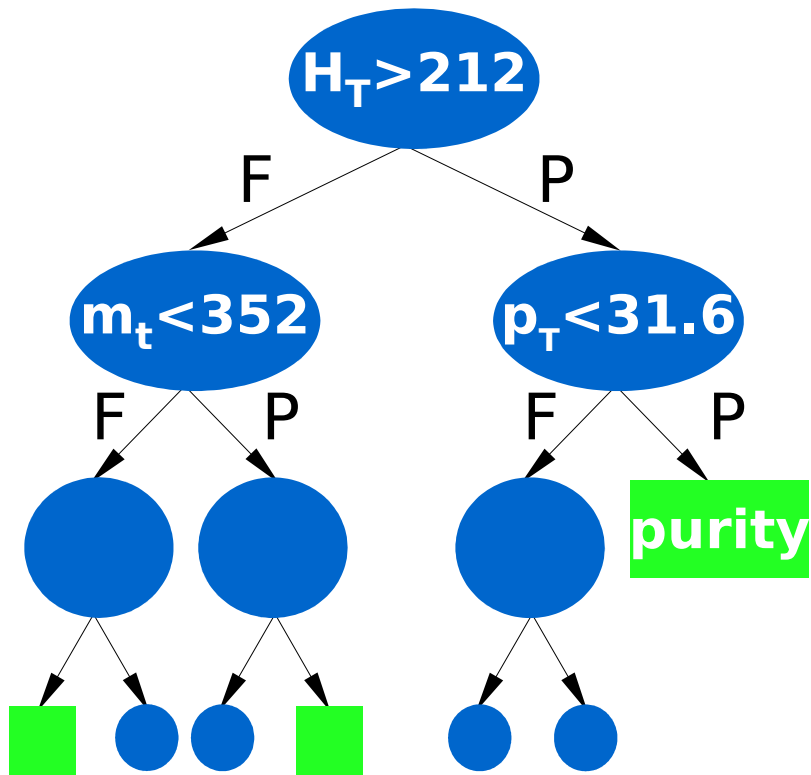
before cuts S/B = 1/52 1/32
after cuts S/B = 1/34 1/21



Cut-based analysis: cuts

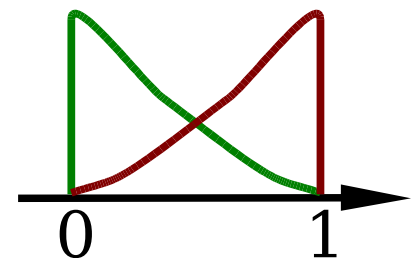
Channel	s-channel		t-channel	
	Variables	Cuts	Variables	Cuts
Electron				
=1 Tag	$p_T(\text{jet1}_{\text{tagged}})$	$> 27 \text{ GeV}$	$H_T(\text{alljets})$	$> 71 \text{ GeV}$
	$M(\text{alljets} - \text{jet1}_{\text{tagged}})$	$< 70 \text{ GeV}$	$M(\text{alljets})$	$> 57 \text{ GeV}$
	$\sqrt{\hat{s}}$	$> 196 \text{ GeV}$	$\sqrt{\hat{s}}$	$> 203 \text{ GeV}$
			$ 175 - M(\text{top}_{\text{tagged}}) $	$< 57 \text{ GeV}$
			$p_T(\text{jet1}_{\text{tagged}})$	$> 21 \text{ GeV}$
≥ 2 Tags	$p_T(\text{jet1}_{\text{tagged}})$	$> 42 \text{ GeV}$	$p_T(\text{jet1}_{\text{tagged}})$	$> 34 \text{ GeV}$
	$M(\text{alljets} - \text{jet1}_{\text{tagged}})$	$< 98 \text{ GeV}$	$M(\text{alljets} - \text{jet1}_{\text{tagged}})$	$< 75 \text{ GeV}$
	$H(\text{alljets} - \text{jet1}_{\text{tagged}})$	$< 304 \text{ GeV}$	$H(\text{alljets} - \text{jet1}_{\text{tagged}})$	$< 504 \text{ GeV}$
	$H(\text{alljets} - \text{jet}_{\text{best}})$	$< 304 \text{ GeV}$	$H(\text{alljets} - \text{jet}_{\text{best}})$	$< 504 \text{ GeV}$
Muon				
=1 Tag	$p_T(\text{jet1}_{\text{tagged}})$	$> 33 \text{ GeV}$	$ 175 - M(\text{top}_{\text{tagged}}) $	$< 60 \text{ GeV}$
	$M(\text{alljets} - \text{jet1}_{\text{tagged}})$	$< 74 \text{ GeV}$	$\sqrt{\hat{s}}$	$> 210 \text{ GeV}$
	$H(\text{alljets} - \text{jet1}_{\text{tagged}})$	$< 504 \text{ GeV}$	$M(\text{alljets})$	$> 70 \text{ GeV}$
	$H(\text{alljets} - \text{jet}_{\text{best}})$	$< 504 \text{ GeV}$	$H_T(\text{alljets})$	$> 58 \text{ GeV}$
≥ 2 Tags	$p_T(\text{jet1}_{\text{tagged}})$	$> 33 \text{ GeV}$	$ 175 - M(\text{top}_{\text{tagged}}) $	$< 213 \text{ GeV}$
	$M(\text{alljets} - \text{jet1}_{\text{tagged}})$	$< 74 \text{ GeV}$		
	$H(\text{alljets} - \text{jet1}_{\text{tagged}})$	$< 504 \text{ GeV}$		
	$H(\text{alljets} - \text{jet}_{\text{best}})$	$< 504 \text{ GeV}$		

Decision Trees

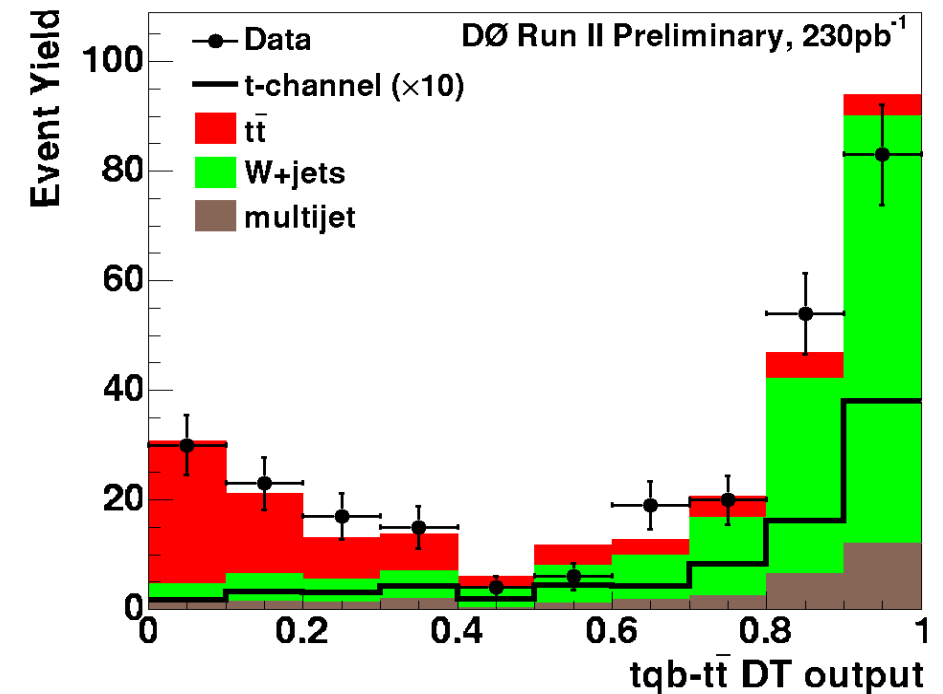
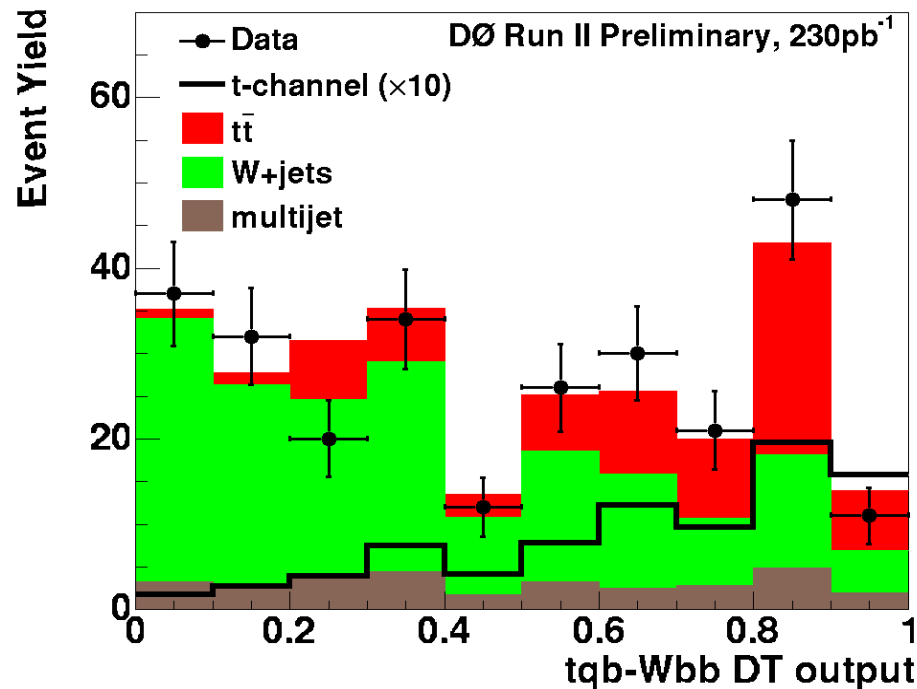
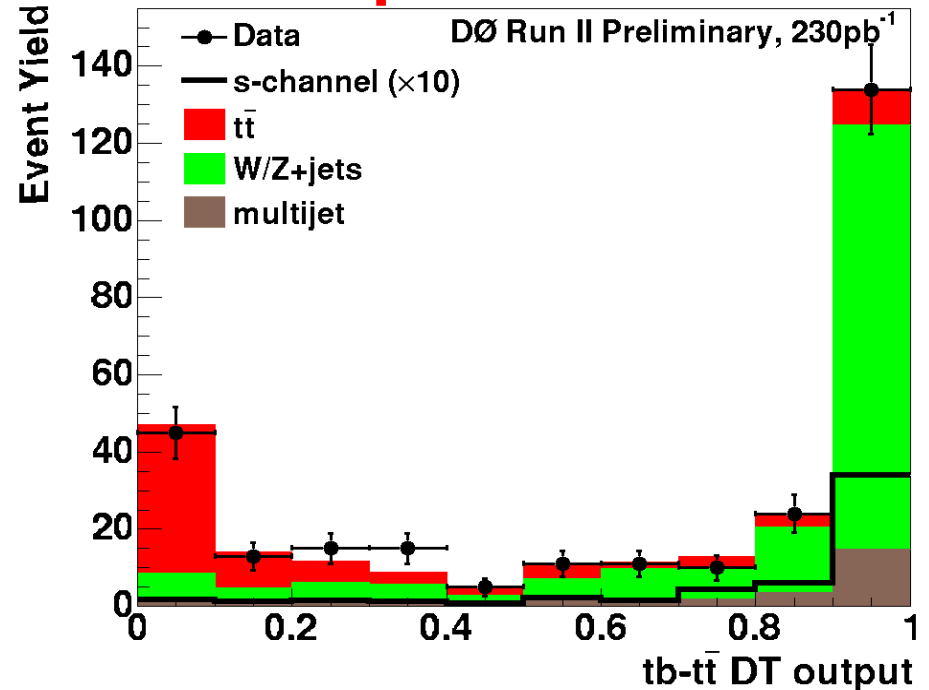
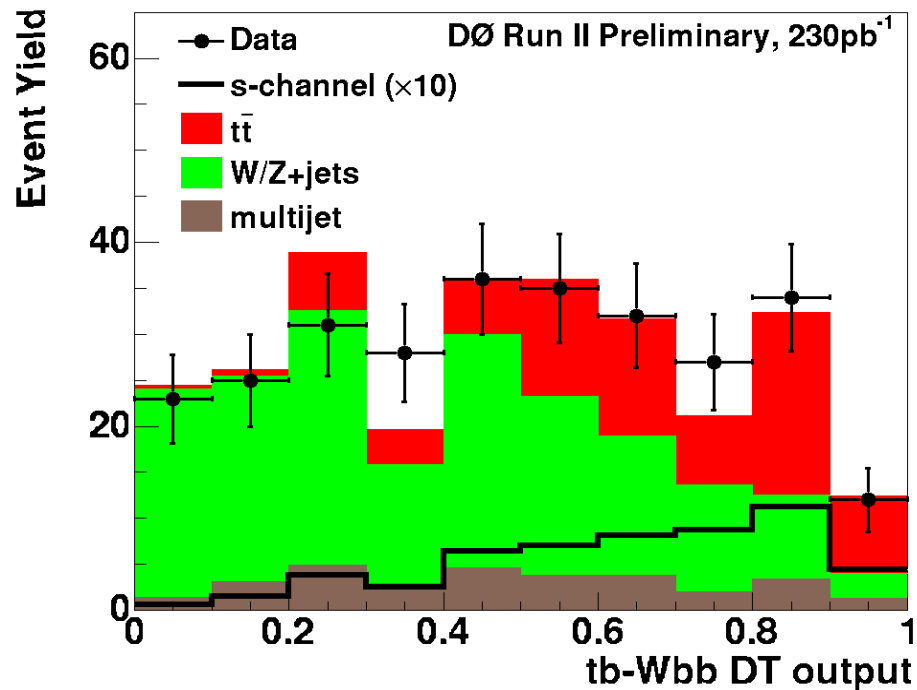
Multivariate technique widely used in social sciences
Recently applied to HEP: MiniBooNE (object ID)
Gives probability for an event to be signal



- ▶ Send each event down the tree
- ▶ Each node  corresponds to a cut
Pass cut (P): right branch
Fail cut (F): left branch
- ▶ A leaf  corresponds to a node without branches
- ▶ Define $\text{purity} = N_S / (N_S + N_B)$
- ▶ Training: optimize Gini improvement
 $\text{Gini} = 2 N_S N_B / (N_S + N_B)$
- ▶ Output:
purity for each event



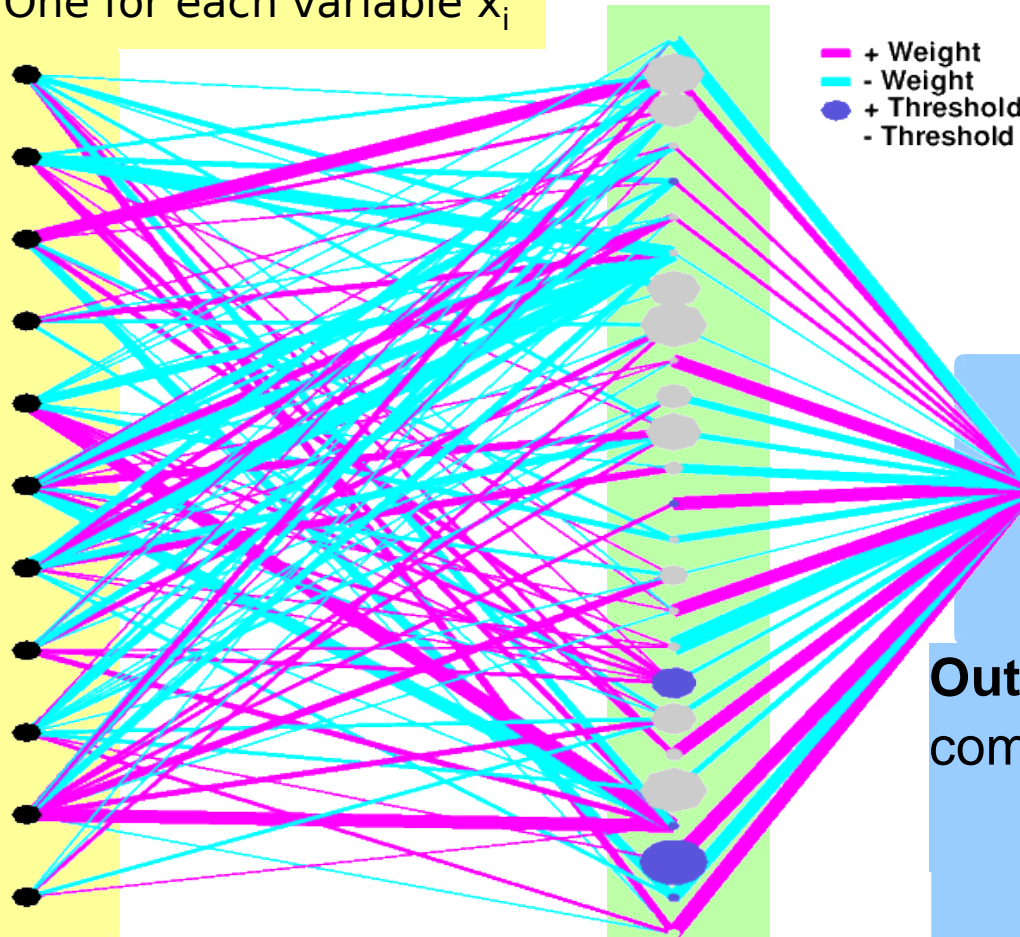
Decision Tree outputs



Neural Networks

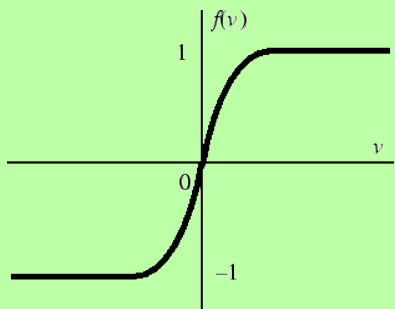
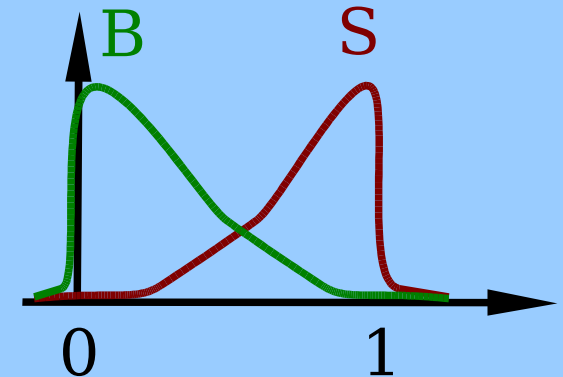
Input Nodes: One for each variable x_i

$M_T(\text{jet1, jet2})$
 $M(\text{alljets})$
 $p_T(\text{jet1, jet2})$
 $p_T(\text{notbest2})$
 $p_T(\text{notbest1})$
 $\cos(\angle, Q(\angle) \times z)_{\text{besttop}}$
 $M(W, \text{best})$
 $M(W, \text{tag1})$
 $\Delta R(\text{jet1, jet2})$
 \sqrt{s}
 $p_T(\text{tag1})$



Output Node: linear combination of hidden nodes

$$f(\vec{x}) = w'_k n_k(\vec{x}, \vec{w}_k)$$



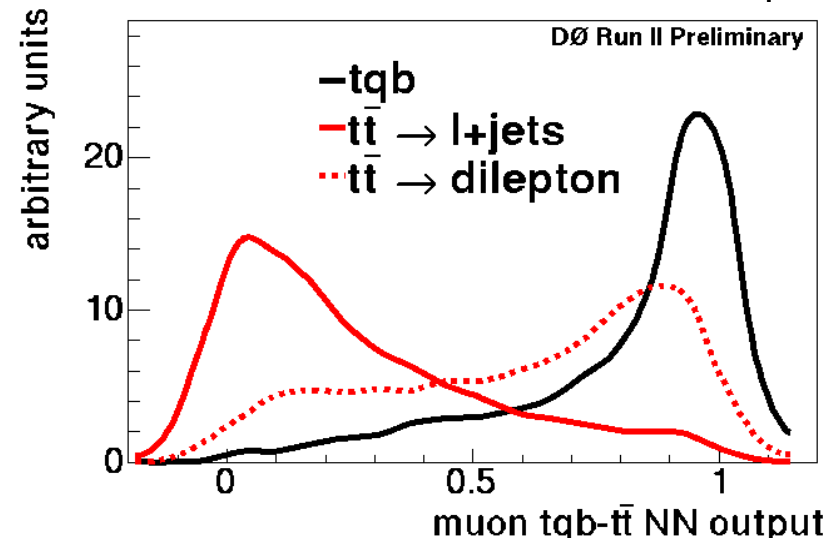
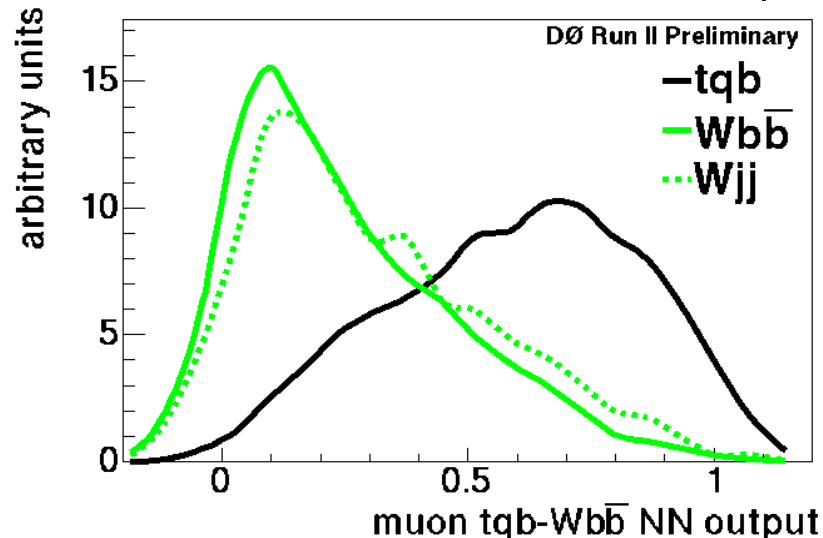
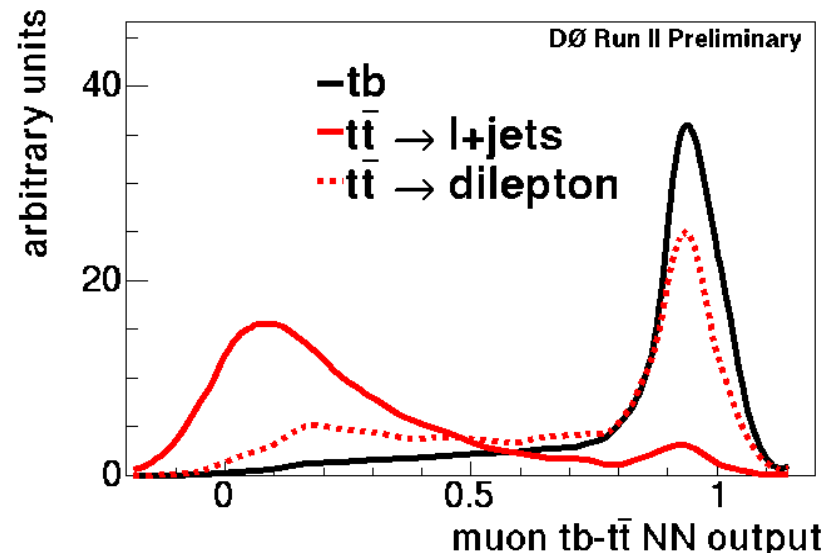
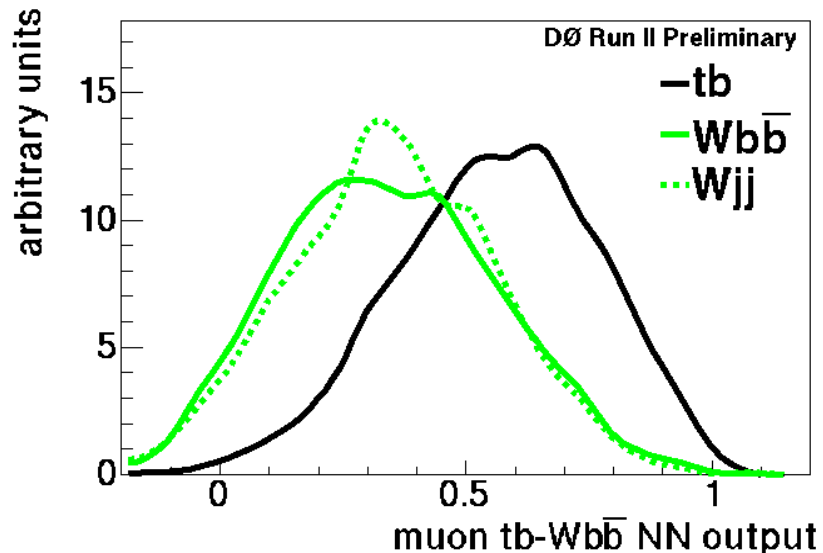
Hidden Nodes: Each is a sigmoid dependent on the input variables

$$n_k(\vec{x}, \vec{w}_k) = \frac{1}{1 + e^{-w_{ik} x_i}}$$

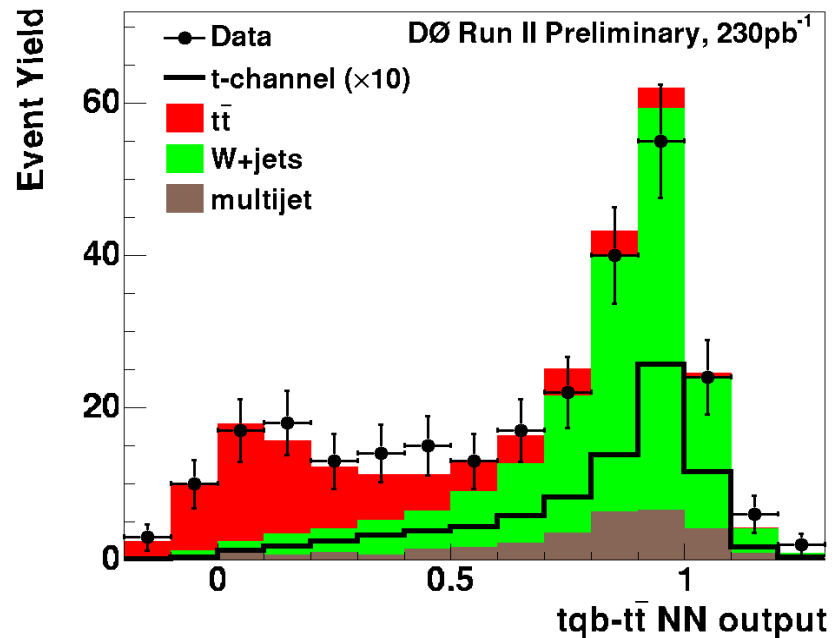
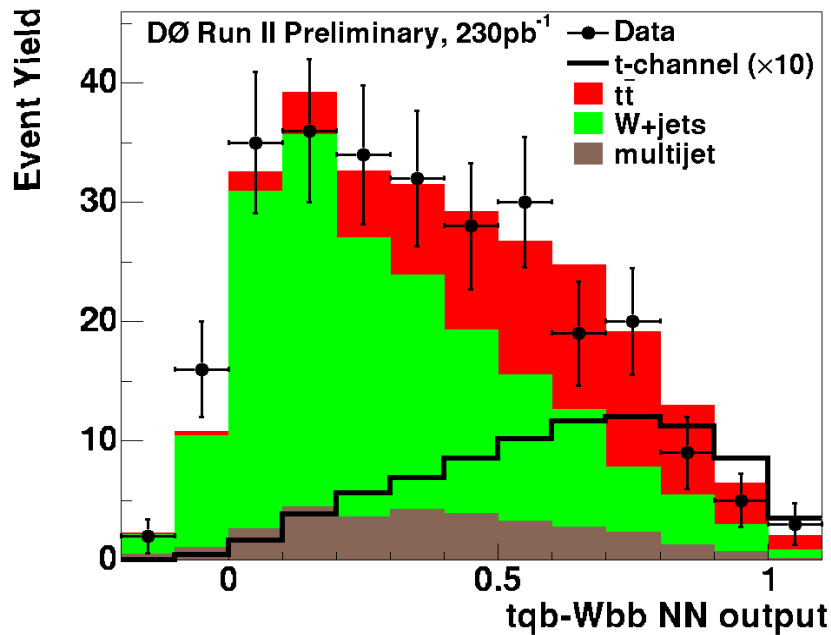
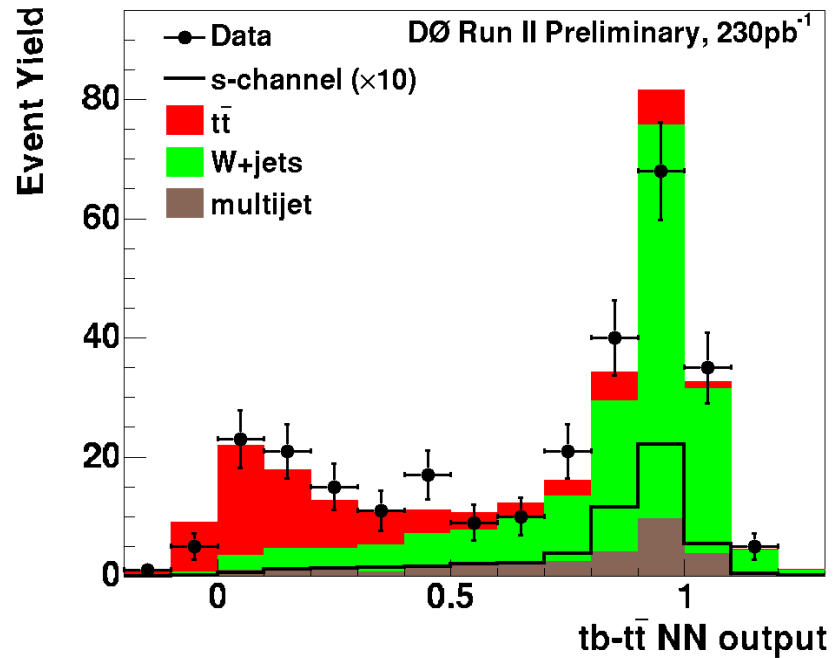
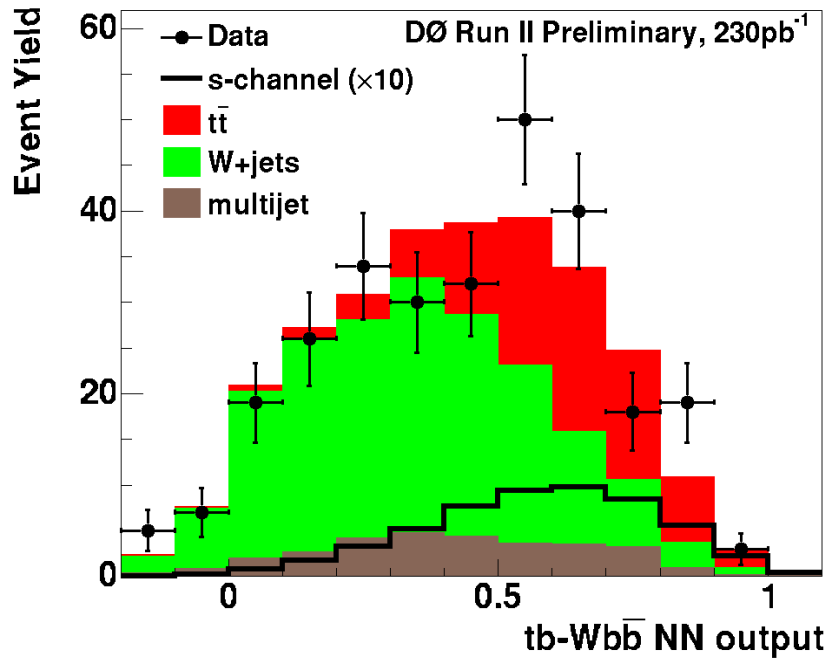
Neural Networks separation

Focus on two main backgrounds:

- ▶ Wbb takes care well of Wjj ($j=g,u,d,s,c$)
- ▶ $tt \rightarrow$ dilepton is small compared to $tt \rightarrow \ell + \text{jets}$



Neural Networks output

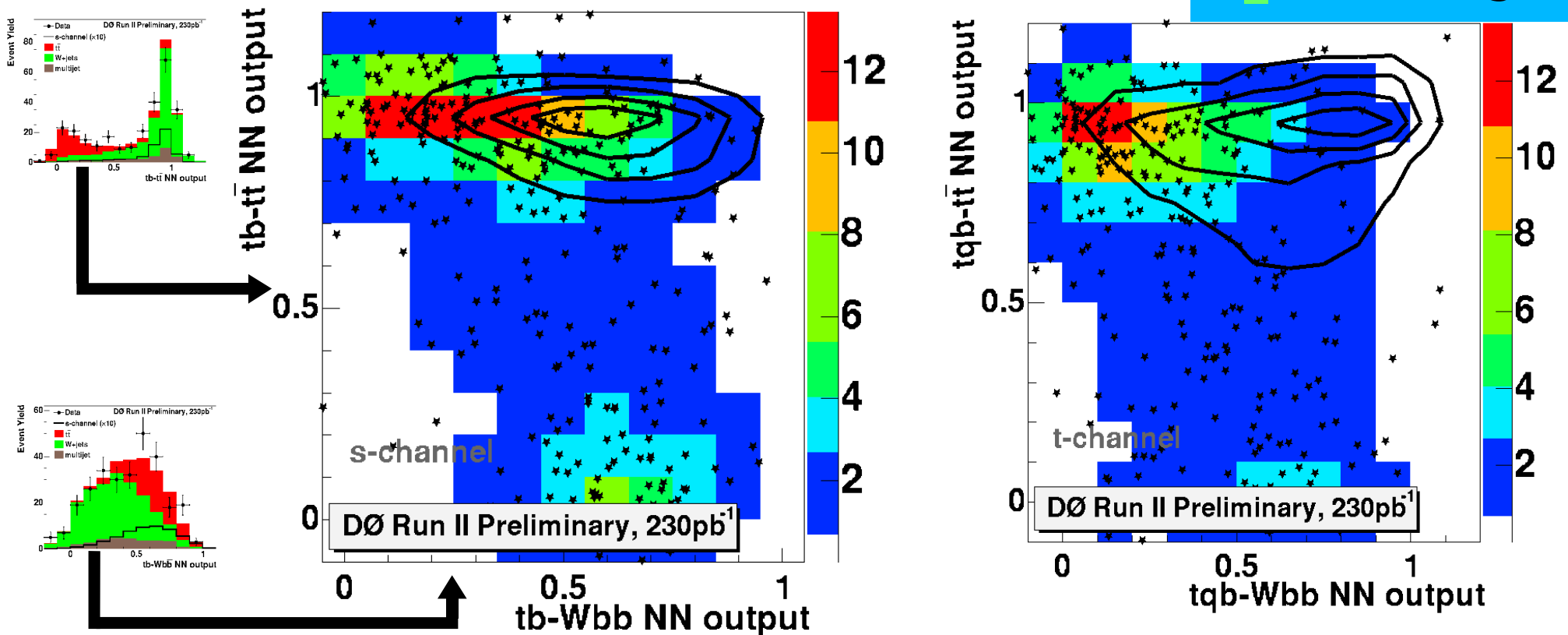
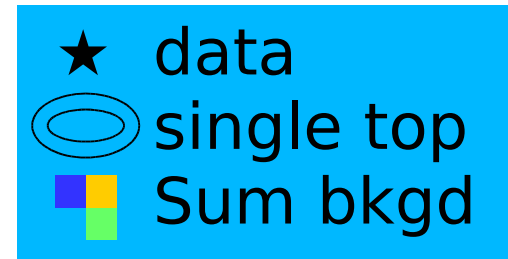


Limits from binned likelihood

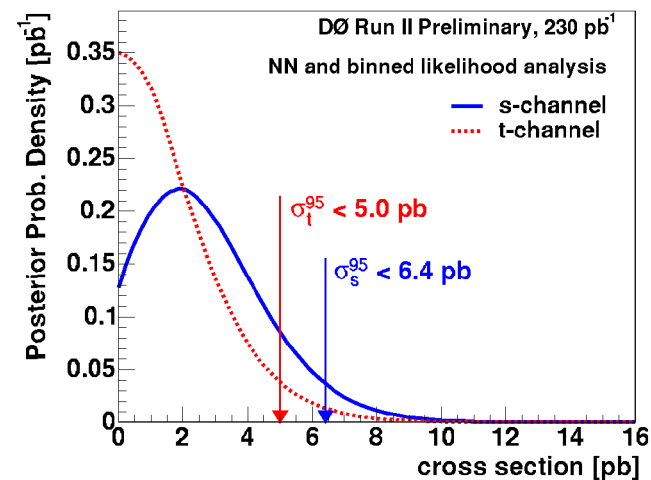
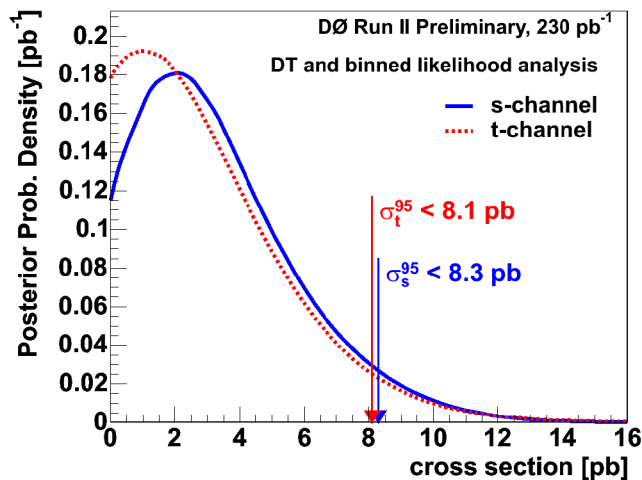
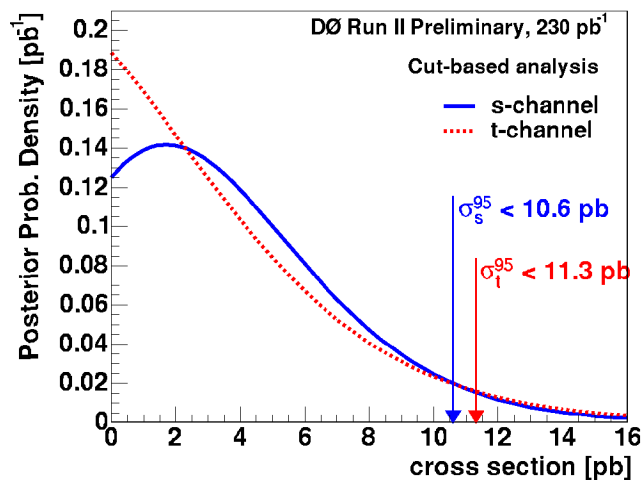
- ▶ No evidence for single top signal
- ▶ Set 95% CL upper cross section limit with Bayesian approach
- ▶ Use 2D histograms as input to build binned likelihood
- ▶ Including bin-by-bin systematics and correlations

Used for DT and NN analyses

Cut-based analysis uses likelihood from event count



DØ results

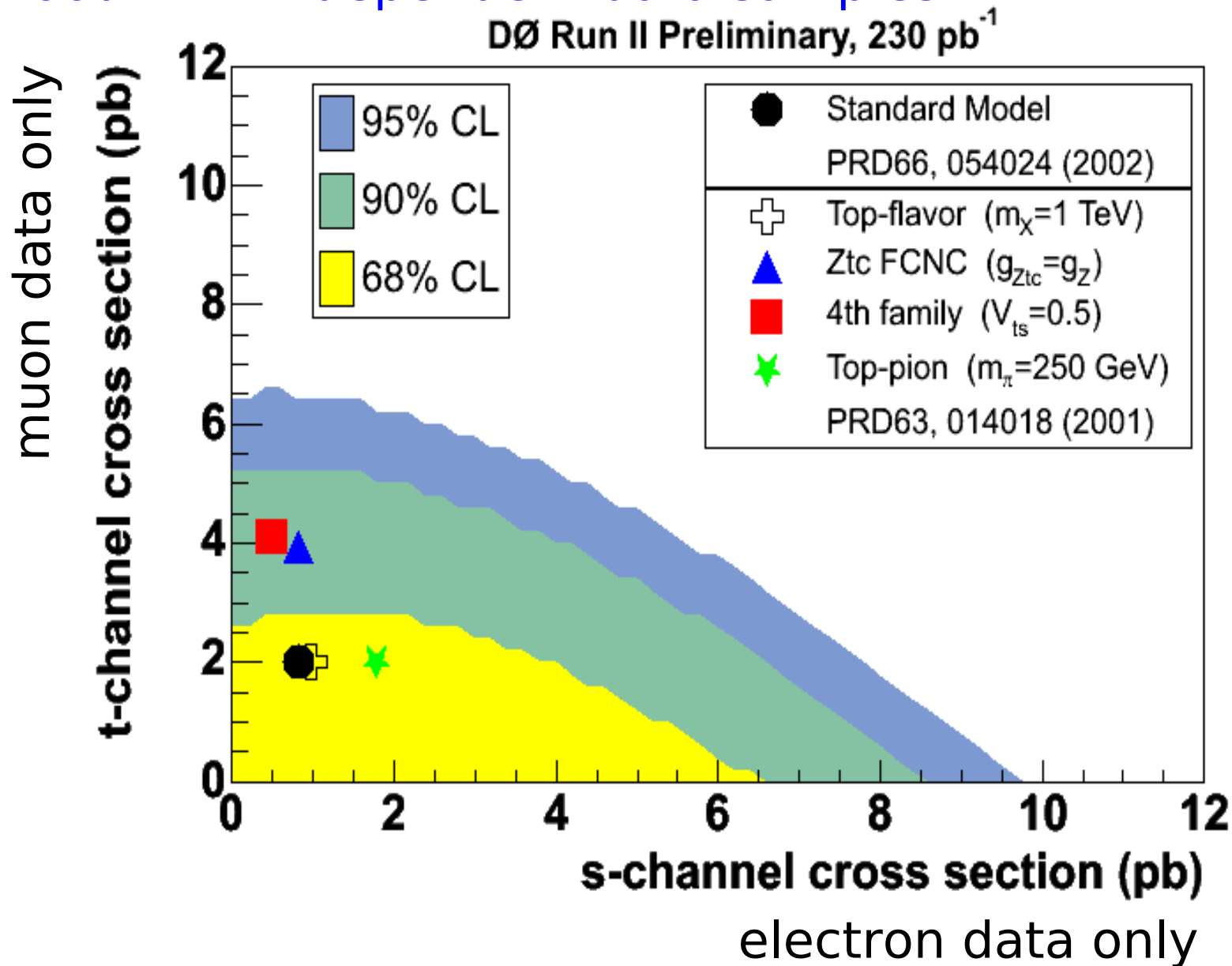


$\mathcal{L} = 230 \text{ pb}^{-1}$	Cut-based		Decision Trees		Neural Networks	
	Observed	Expected	Observed	Expected	Observed	Expected
s-channel	10.6	9.8	8.3	4.5	6.4	4.5
t-channel	11.3	12.4	8.1	6.4	5.0	5.8

- ▶ Use Bayesian approach to combine channels (e, μ and 1 tag, 2 tags)
- ▶ Take systematics and correlations into account
- ▶ Decision Trees and Neural Networks have similar sensitivity
- ▶ Multivariate analysis + shape information from output:
→ **factor 2 better than simple cuts**

Beyond the SM

$D\emptyset$ exclusion contours on the (σ_s, σ_t) plane, built from a joint likelihood with independent data samples



Summary and next steps

95% CL Measured Upper Limits in pb

	<i>s</i> -channel	<i>t</i> -channel
DØ Run I, 90 pb ⁻¹	17	22
CDF Run II, 162 pb ⁻¹	13.6	10.1
DØ Run II , 230 pb ⁻¹ cuts	10.6	11.3
DTs & binned likelihood	8.3	8.1
NNs & binned likelihood	6.4	5.0
NLO theory	=0.88	=1.98

Current analyses need a few fb⁻¹ for observation



Upgrade to improved b-tagging

Explore multivariate methods:

Kinematic fitter, Neural Network, Matrix Element, optimized likelihoods



Optimizing current analyses

Increasing acceptance

Improving object ID

New methods under study

Increased dataset!

Conclusions

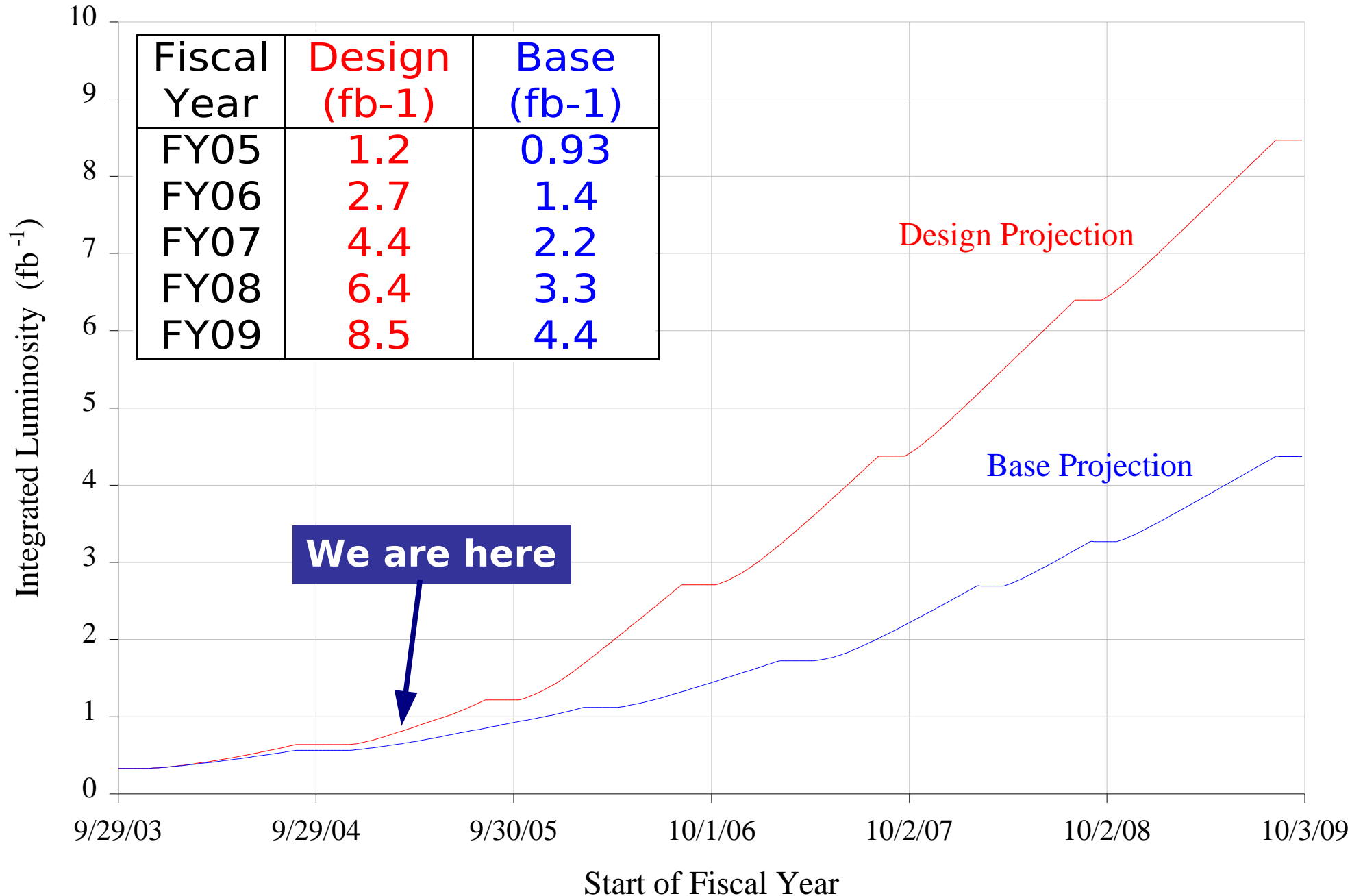
- ▶ Single top search at the Tevatron is more challenging than anticipated (see Matt Bowen's talk next)
- ▶ Signal generators are well established
- ▶ W+jets & heavy flavor fractions estimation are the big issues
- ▶ Need advanced analysis techniques to extract signal, but these require to understand well the systematics

- ▶ CDF has published the first Run II limits
- ▶ DØ is about to submit the results presented here
- ▶ Reaching sensitivity to new Physics

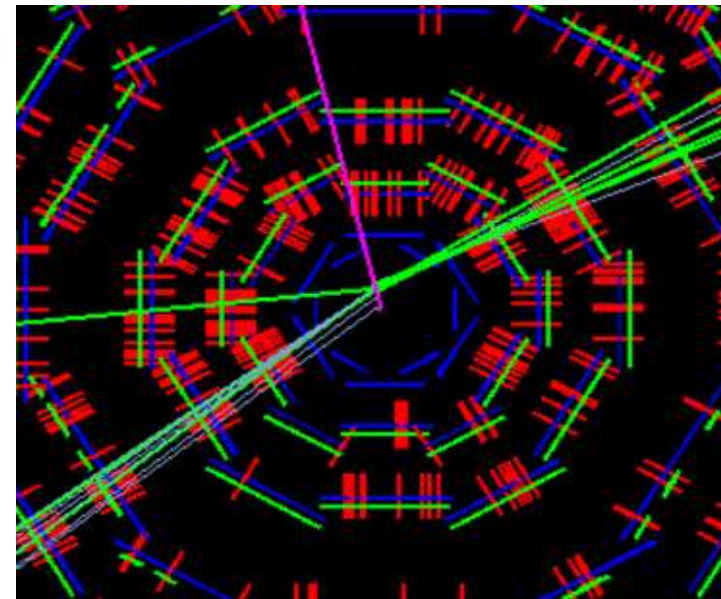
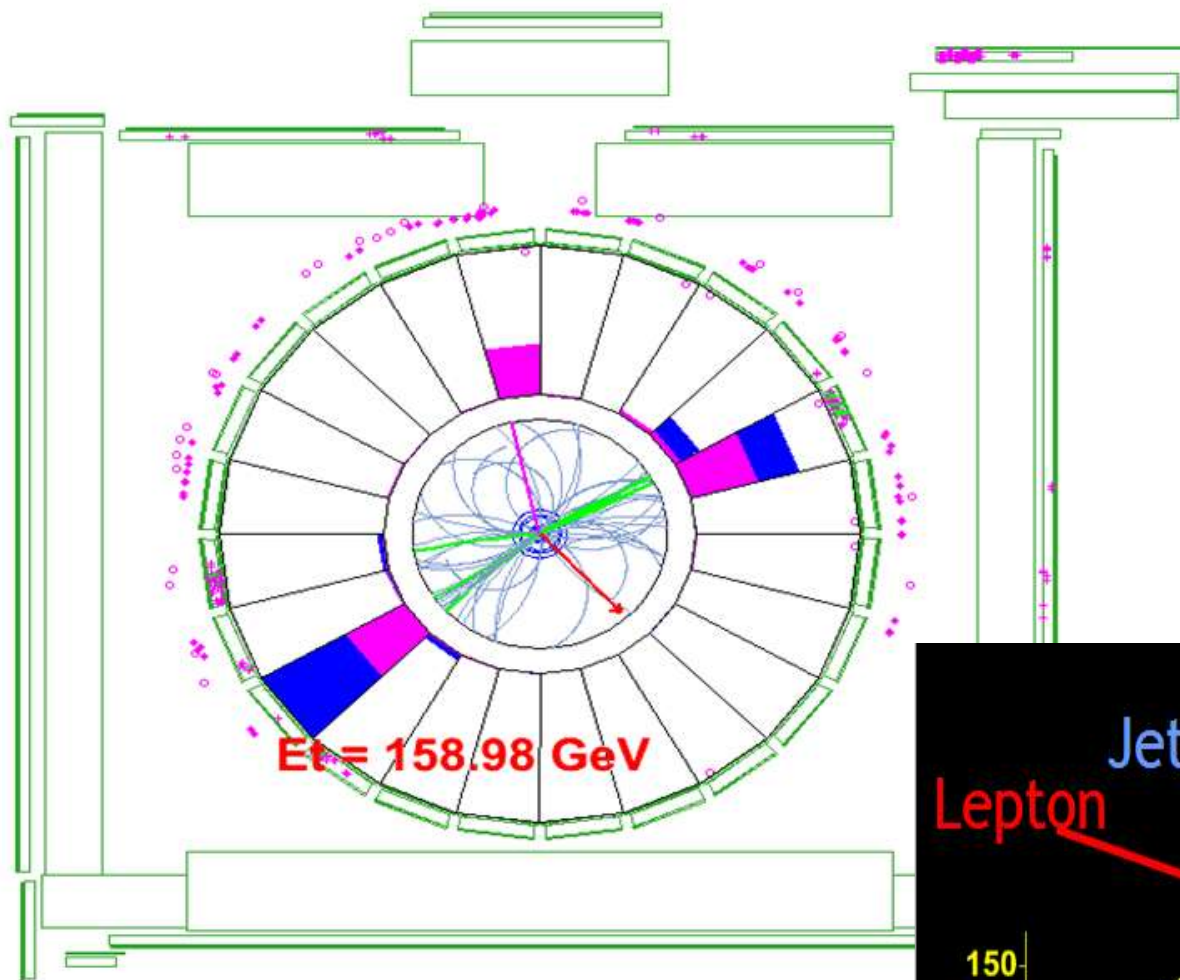
- ▶ Plenty of room for improvements:
 - ▶ New analysis methods
 - ▶ Better object ID, better b-tagging
 - ▶ More data helps reduce the systematics

Our detectors are being understood better every day
Data is pouring in (expect $\times 3$ dataset by the end of the year)
This is an exciting time for single top

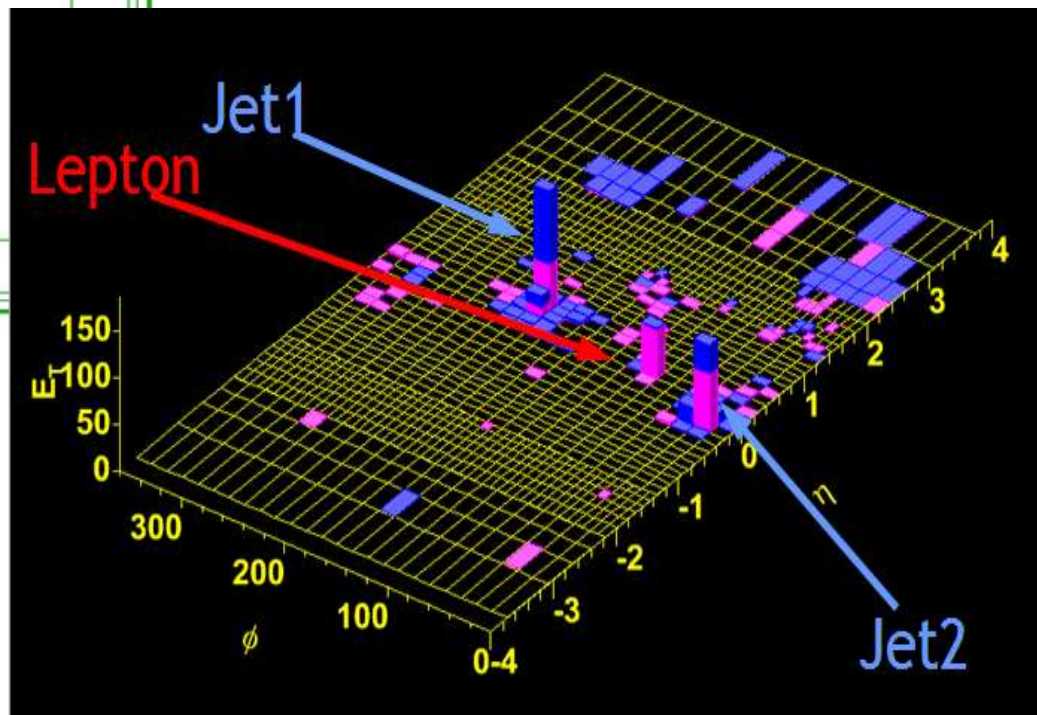
Tevatron luminosity prospects



Extra1: CDF candidate event



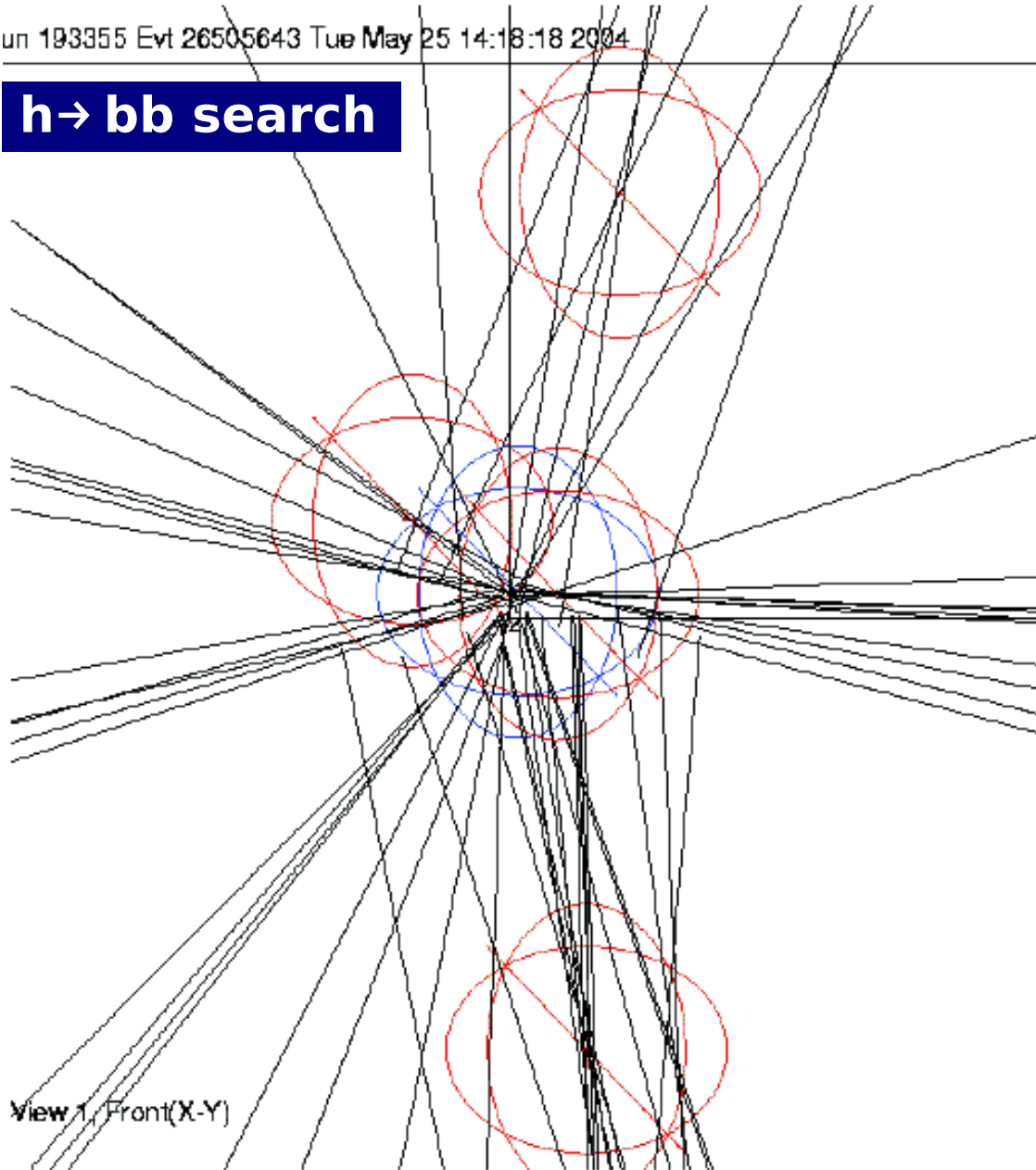
Run 153389 Event 361345
CEM electron $E_T = 50.9 \text{ GeV}$
MET = 27.4 GeV
Jet1 $E_T = 175.2 \text{ GeV}$, $\eta = 0.45$
Jet2 $E_T = 147.2 \text{ GeV}$, $\eta = -0.13$



Extra2: quadr. b-tagged event in DØ

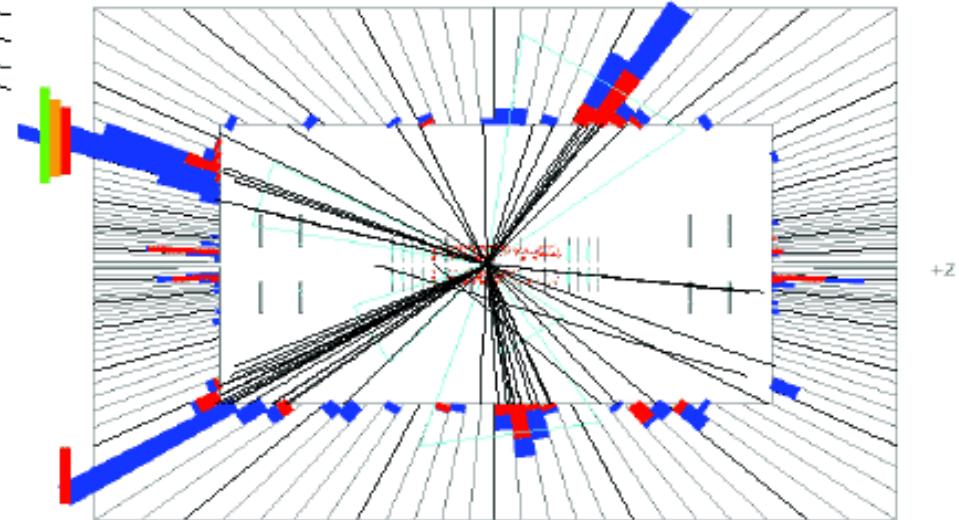
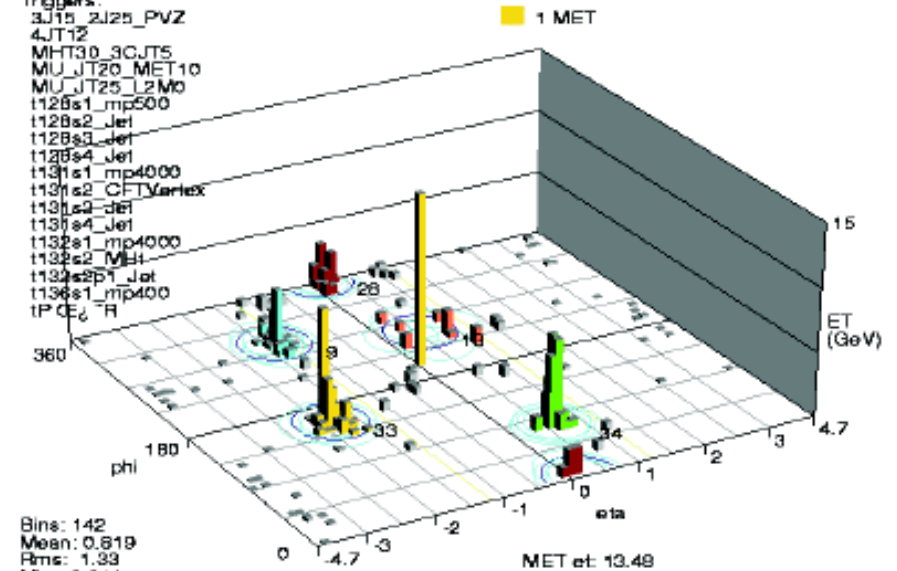
Run 193355 Evt 26505643 Tue May 25 14:18:18 2004

h → bb search



Run 193355 Evt 26505643 Tue May 25 14:18:18 2004

Triggers:
3J15_2J25_PVZ
4JT12
MHT30_3CJTS
MU_JT20_MET10
MU_JT25_L2Mo
t128s1_mp500
t128s2_Jet
t128s3_Jet
t128s4_Jet
t131s1_mp4000
t131s2_CFTVertex
t131s2-Jet
t131s4_Jet
t132s1_mp4000
t132s2_MHT
t132s251_Jet
t136s1_mp400
TPCE_R



Extra 3: More info

CDF: <http://www-cdf.fnal.gov/physics/new/top/top.html>

DØ: <http://www-d0.fnal.gov/Run2Physics/WWW/results/top.htm>